

Cyclical Budgetary Policy and Economic Growth: What Do We Learn from OECD Panel Data?*

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Abstract: This paper uses yearly panel data on OECD countries to analyze the relationship between growth and the cyclical nature of government debt. We develop new time-varying estimates of the cyclical nature of public debt. Our main findings can be summarized as follows: (i) less procyclical public debt growth can have significantly positive effects on productivity growth, in particular when financial development is lower; (ii) public debt growth has become increasingly countercyclical in most OECD countries over the past twenty years, but this trend has been less pronounced in the EMU; (iii) less financially developed or more open economies display less countercyclical public debt growth.

1 Introduction

A common view among macroeconomists, is that there exists a perfect dichotomy between macroeconomic policy (budget deficit, taxation, money supply) which should affect primarily the short-run, and long-run economic growth which, if anything, should depend only upon structural characteristics of the economy (property right enforcement, market structure, market mobility and so forth). That macroeconomic policy should not be a key source of growth, is further hinted at by recent contributions such as Acemoglu et al (2004) and Easterly (2005), which argue that the correlation between macroeconomic volatility and growth (Acemoglu et al) or those between growth and macroeconomic variables (Easterly), become insignificant once one controls for institutions.

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In this paper we question that view by showing that, the cyclicity of public debt growth is significantly correlated with GDP growth, with more a countercyclical public debt policy being more growth-enhancing the lower the country's level of financial development. These results hold in a sample of OECD countries with comparable institutional environments.

Our contribution in this paper is three-fold. It is first to compute and analyze the cyclicity of government debt on a panel of OECD countries, that is, how government debt responds to fluctuations in the output gap over time. Second, it is to use these yearly panel data to assess the importance for growth of moving towards more countercyclical budgetary policies at various levels of financial development. Third, it is to investigate some determinants of the procyclicality of public debt. Our main findings can be summarized as follows: (i) public debt growth has become increasingly countercyclical in most OECD countries over the past twenty years, but this trend has been significantly less pronounced in the EMU; (ii) less procyclical public debt growth can have significantly positive effects on growth when financial development is lower; in particular our estimates suggest that the eurozone could increase its annual growth rate by 0.15 percentage points per year by making its public debt growth become as countercyclical as that in the US; (iii) less financially developed or more open economies display less countercyclical government debt growth.

The idea that cyclical macroeconomic policy should affect productivity growth, is more in line with the Schumpeterian view of business cycles and growth, whereby recessions provide a cleansing mechanism for correcting organizational inefficiencies and for encouraging firms to reorganize, innovate or reallocate to new markets. The cleansing effect of recessions is also to eliminate those firms that are unable to reorganize or innovate. Now, if firms could always borrow enough funds to either reorganize their activities or move to new activities and markets, and the same was true for workers trying to relocate from one job to another, the best would be to recommend that governments do not intervene over the business cycle, and instead let markets operate.

However, suppose that the borrowing capacity of firms is proportional to their current earnings (the factor of proportionality is what we refer to as the credit multiplier, with a higher multiplier reflecting a higher degree of financial development in the economy). In a recession, current earnings are reduced, and therefore so is the firms' ability to borrow in order to make new innovative investments or simply maintain previous innovation programs in the face of idiosyncratic liquidity shocks. This, in turn suggests that a countercyclical budgetary policy may foster innovation and growth by reducing the negative consequences of a recession (or a bad aggregate shock) on firms' innovative investments. For example, the government may decide to increase the volume of its public investments or that of public consumption, thereby fostering the demand for private firms' products. Or the government may choose to directly increase its subsidies to private enterprises, thereby increasing their liquidity holdings and thus making it easier for them to face idiosyncratic liquidity shocks without having to sacrifice R&D or other types of longer-term growth-enhancing investments. A natural implication of the above argument is that the lower the level of

financial development, that is, the tighter the credit constraints faced by firms, the more growth-enhancing such countercyclical policies should be, a conclusion at odds with the dichotomous view that prevails among macroeconomists.

While we do not know of any previous attempt at analyzing the growth effects of countercyclical budgetary policies, analyses of the determinants of the cyclicity of budgetary policies already exist in the literature. For example, Alesina and Tabellini (2005) argue that more corrupt democracies will tend to run more procyclical fiscal policy. The idea is that, in good times, voters demand that the government cut taxes or provide more public services instead of reducing debt, because they cannot observe the debt reduction and can suspect the government of appropriating the rents associated with good economic conditions. In equilibrium, this leads to a more procyclical policy as the moral hazard problem worsens, in the sense that governments are more likely to divert public resources in booms. They also show that this mechanism tends to be more powerful in explaining the variation observed in the data than borrowing constraints alone. While Alesina and Tabellini (2005) are using a large sample of countries and explore cross-section variations, in this study we use panel analysis on OECD countries. This makes the use of corruption indices impractical for two reasons. First, there is almost no cross-sectional variation in corruption indices within the OECD. Second, there is even less variation of these indices across time for individual countries.

In a similar vein, Calderon et al. (2004) show that emerging market economies with better institutions are more able to conduct a countercyclical fiscal policy¹. Their empirical analysis is based on the International Country Risk Guide. Although the variation in this indicator is limited across OECD countries and time, it presents somewhat more variation than corruption indexes².

Other papers such as Gali and Perotti (2003) and Lane (2003) focus like we do on OECD countries. Gali and Perotti investigate whether fiscal policy in the European Monetary Union (EMU) has become more procyclical after the Maastricht treaty. They find no evidence for such a development. They do find however that while there is a trend in the OECD towards a more countercyclical fiscal policy over time, the EMU is lagging behind that trend. Lane (2003) is probably the paper that comes closer to the analysis developed in the fourth section of our paper. Lane examines the cyclical behavior of fiscal policy within the OECD. He then uses trade openness, output volatility, output per capita, the size of the public sector and a index for political power dispersion to examine cross-country differences in cyclicity. The reason why

¹There is also the paper by Talvi and Vegh (2000), where it is argued that high output volatility is most likely to generate a procyclical government spending. The idea is that running a budget surplus generates political pressures to spend more: the government therefore minimizes that surplus and becomes pro-cyclical. This movement is then accentuated by a volatile output, and therefore a volatile tax base.

²We have also used these indicators in our analysis. However, they typically have no significant effect on GDP growth over time in our sample. Moreover, as they are less widely available than our main variables of interest, their use considerably restricts the available sample, leading to less precise estimates. Finally, we have therefore decided not to use these indicators in the results reported here.

power dispersion may play a role is taken from Lane and Tornell (1998): when multiple political groups compete for public spending, the latter may become more procyclical. Indeed, no group wants to let any substantial fiscal surplus subsist because they are afraid that this will not lead to debt repayment but to other groups appropriating that surplus. Lane finds in particular evidence that GDP growth volatility, trade openness and political divisions lead to a more procyclical spending pattern, even though the effect of political divisions is not present for all categories of spending. We contribute to this literature by using yearly panel data to analyze the cyclicity of budgetary policies and its determinants within OECD countries, and we show that the degree of financial development is an important element to explain both, cross-country and within country variations in such policies, while future or present EMU membership explains cross-country variations.

Most closely related to our second stage analysis of the effect of countercyclical budgetary policy on growth, are Aghion-Angeletos-Banerjee-Manova (2005), henceforth AABM, and Aghion-Bacchetta-Ranciere-Rogoff (2006), henceforth ABRR. AABM develop a model to explain why macroeconomic volatility is more negatively correlated with productivity growth, the lower financial development, and they test this prediction using cross-country panel data. ABRR move from a closed real to an open monetary economy and show that a fixed nominal exchange rate regime or lower real exchange rate volatility are more positively correlated with productivity growth, the lower financial development and the lower the ratio of real shocks to financial shocks.

The remaining part of the paper is organized as follows. Section 2 develops the first stage analysis of the cyclicity of public debt growth for each OECD country and each year covered by our panel data set. In Section 3, we regress GDP growth on financial development, the cyclicity coefficients computed in the first-stage regressions, and the interaction between the two. In Section 4, we uncover some main determinants of the cyclicity of public debt. Finally, Section 5 concludes.

2 First stage regressions: the cyclicity of public spending in the cross-country panel

2.1 Data

Panel data on GDP, the GDP gap (*ygap*), the GDP deflator, government gross debt (*ggfl*), total government disbursements (*ypgt*), government investment (*igaa*) and government consumption (*cgaa*), are taken from the OECD Economic Outlook annual series³. Total government disbursements include government investment, government consumption, debt repayment, subsidies to the

³Codes in parenthesis indicate the names of variables in the dataset. Full documentation available at http://www.oecd.org/findDocument/0,2350,en_2649_34573_1_119669_1_1_1,00.html. Data can be downloaded from sourceoecd.org for subscribers to that service.

private sector, social security and other related transfers, capital transfers and government consumption of fixed capital. Note that debt and other government data refer to general government. Financial development is measured by the ratio of private credit to GDP, and annual cross-country data for this measure of financial development can be drawn from the Levine database⁴. In this latter measure, private credit is all credit to private agents and therefore includes credit to households. The average years of education in the population over 25 years old series is directly borrowed from the Barro-Lee dataset; this measure is only available every five years and has been linearly interpolated to obtain a yearly series. The openness variable is defined as exports and imports over GDP and data on it come from the Penn World Tables 6.1. The population growth, government share of GDP, investment share of GDP also come from the Penn World Tables 6.1. All nominal variables are deflated using the GDP deflator. Summary statistics can be found in Table 1. The sample is an unbalanced panel including the following countries: Australia, Austria, Belgium, Canada, Denmark, Spain, Finland, France, United Kingdom, Germany⁵, Greece, Iceland, Italy, Japan, Netherlands, Norway, New Zealand, Portugal, Sweden, USA.

2.2 Rationale for the specification of the first stage

The baseline model for public debt growth as a function of the output gap, comes from the tax-rate-smoothing model by Barro (1979). In this framework, deficits emerge from temporary deviations of government expenditure from “normal” and from temporary deviations of the tax base, assumed to be represented by real GDP, from “normal.” Since tax-rate smoothing relates to the ratio of public debt to GDP, an interaction of the level of debt with anticipated growth of GDP also factors into budget deficits. Moreover, given the way that real deficits are usually calculated in the national accounts (corresponding to changes in nominal debt divided by a price index), it is the growth of nominal GDP that matters. That is, anticipated inflation influences the “real” deficit.

We assume that the relevant tax base is proportional to real GDP, y_t . We assume further that smoothing of the relevant marginal tax rates (for example, on labor income or consumption or value added) corresponds to smoothing the average tax rates, T_t/y_t , where T_t is real taxes collected in year t .

Let g_t be real government expenditure on purchases and transfers. Suppose that $\log(g_t)$ deviates temporarily from its trend, $[\log(g_t)]^*$. Formally, the trend should correspond to the expected present value of expenditure. In practice, we use an H-P filter to estimate the trend in $\log(g_t)$. The deviation, $\log(g_t) - [\log(g_t)]^*$, is the proportionate departure of g_t from normal. Multiplying by the trend or normal value, $(g_t)^*$, gives the amount of real debt issue required to

⁴Data downloadable at http://www.econ.brown.edu/fac/Ross_Levine/Publication/CrossCountryFinStructure.zip.

⁵All level variables are adjusted for the German reunification. The adjustment involves regressing each variable of interest on time and a constant in the ten years before 1991 (data based on West Germany only). We then use the estimated coefficients to predict the values for 1991 to 2000. We take the average ratio between actual and predicted values in the years 1991 to 2000. We use this ratio to proportionally adjust values before 1991.

finance temporary expenditure (rather than having temporarily high tax rates).

Suppose that $\log(y_t)$ deviates temporarily from its trend, $[\log(y_t)]^*$. A positive value corresponds to a boom and a negative one to a recession. Again, we use an H-P filter (or, alternatively, measures of capacity output) to calculate $[\log(y_t)]^*$. Given the behavior of g_t , tax-rate smoothing implies that a temporary excess of $\log(y_t)$ from $[\log(y_t)]^*$ calls for an equi-proportionate excess of real taxes, T_t , from normal. Normal real taxes correspond to normal or trend expenditure, $(g_t)^*$. Therefore, the product of $\log(y_t) - [\log(y_t)]^*$ and $(g_t)^*$ gives the budget surplus (corresponding to a temporarily high level of real taxes collected) associated with a boom.

Given $\log(g_t) - [\log(g_t)]^*$ and $\log(y_t) - [\log(y_t)]^*$, tax-rate smoothing calls for expanding the level of real debt, b_t , along with expansions of real GDP, y_t . That is, if $\log(g_t) = [\log(g_t)]^*$ and $\log(y_t) = [\log(y_t)]^*$, the debt-GDP ratio should stay constant. Therefore, the change in the real debt, $b_t - b_{t-1}$, includes a term γb_{t-1} , where γ is the (trend) growth rate of real GDP.

The national accounts typically measure the real budget deficit as the real value of the change in the nominal debt (because nominal government expenditure includes interest payments computed from the nominal interest rate). When measured this way, tax-rate smoothing implies that the real budget deficit includes another term, πb_{t-1} , where π is the (expected) inflation rate. That is, the measured real budget deficit depends on the overall term $(\gamma + \pi)b_{t-1}$, where $\gamma + \pi$ is the growth rate of nominal GDP.

The term $(\gamma + \pi)b_{t-1}$ should move closely with the real value of nominal interest payments. The difference is that nominal interest payments depend on the real interest rate, r , rather than the growth rate of real GDP, γ . If we generate a dependent variable by subtracting the real value of nominal interest payments from the measured real budget deficit, the coefficient on the variable b_{t-1} on the right-hand side should be $\gamma - r$, which we treat as a constant. This constant would be negative in the standard deterministic model. (However, with uncertainty, the real rate r on government debt could be smaller than γ , the mean growth rate of real GDP.)

The baseline tax-smoothing model has no tendency for the debt-GDP ratio to revert to a stationary mean, such as zero. (More generally, the ratio might revert to something positive, possibly dependent on other assets held by the government.) If there were a tendency for the debt-GDP ratio to revert toward zero, we might pick up this effect from the coefficient on the stock of real debt, b_{t-1} . Thus, a negative coefficient on b_{t-1} could represent this mean reversion, along with the effect $\gamma - r$ already mentioned.

Our empirical counterpart of the tax-rate smoothing model of budget deficits is then for each country i :

$$\frac{(b_{it} - b_{i,t-1}) - i_{it}}{y_{it}} = a_{1it} y_{gap,it} \frac{\overline{g_{it}}}{y_{it}} \quad (1)$$

$$+ a_{2it} \{ \ln(g_{it}) - \overline{\ln(g_{it})} \} \frac{\overline{g_{it}}}{y_{it}} + a_{3it} \frac{b_{i,t-1}}{y_{it}} + a_{4it} + \varepsilon_{it}$$

where $\varepsilon_{it} \sim N(0, \sigma_\varepsilon^2)$.

Empirically, the variables are defined as follows:

- b_{it} : gross government debt in country i at year t
- y_{it} : the GDP in country i and year t
- $y_{gap,it}$: the GDP gap in country i and year t as computed by the OECD based on a production function approach.
- i_{it} : interest payments made by the government in country i and year t
- g_{it} : total government disbursements in country i and year t
- ε_{it} : error term

The a coefficients to be estimated are for the purpose of this paper, assumed to be potentially time-varying, which is why we write a_{jit} to denote the coefficient on the variable j in country i at time t .

A bar above a variable indicates that one takes the prediction for this variable using the Hodrick-Prescott filter. A lambda parameter of 25 was chosen, following OECD(1995). The prediction was then computed separately for each country.

Note that $b_{it} - b_{i,t-1}$ is exactly equal to the opposite of the budget balance, so that our left-hand side variable is very close to the opposite of the budget balance as a share of GDP.

With tax smoothing, the predicted coefficients are $a_{1it} = -1$ and $a_{2it} = 1$. The coefficient a_{3it} corresponds, as discussed, to $\gamma - r$ (plus a possible negative effect associated with reversion of the debt-GDP ratio toward zero or some other positive target value).

Even if a government does not precisely pursue tax-rate smoothing, the formulation in equation (1) is useful in the sense that the deviations of the estimated coefficients a from the values prescribed under tax smoothing are informative. For example, if $a_{1it} < -1$, the government is pursuing a more counter-cyclical deficit policy than called for by tax smoothing, and vice versa if $a_{1it} > -1$. A procyclical deficit policy, $a_{1it} > 0$, is very far from tax-rate smoothing; that is, $a_{1it} = 0$ is not the natural baseline. We also examine how the extent of counter-cyclical deficit policy varies over time within countries, as well as across countries.

Finally, when briefly analyzing the cyclicity of government investment or consumption⁶, we just replace government gross debt by these on the left-hand side of equation 1; in such specifications, government interest payments are not subtracted on the left-hand side.

2.3 Econometric methods

Regression based approaches to measure the cyclicity of fiscal policies are now common in the literature and can be found for example in Lane (2003) and Alesina and Tabellini (2005). However, the methods used in these papers give rise to only one observation of cyclicity per country. In order to make full use of the panel structure of our data, we compute instead for each country yearly measures for the cyclicity of different categories of public spending. A first method, which we do not emphasize in the text, estimates time-varying coefficients by using 10-years rolling window ordinary least squares estimates, a straightforward but very noisy and therefore less reliable method. The second method uses local Gaussian-weighted ordinary least squares estimates (also called kernel-based nonparametric regression or local smoothing): for each year, points that are closer in time are given more weight than points that are further away. The third method is to compute time-varying coefficients in the above equation 1 under the assumption that these coefficients follow an AR(1) process.

Consider first the ten-year rolling window OLS method: if $v_{i\tau}$ denotes the value of the budgetary variable under consideration (growth of government debt, government investment, or government consumption as defined above) in country i and year τ , then the method simply amounts to estimating the procyclicality of that variable at year t in country i by running the following regression for each country i , and all possible τ :

$$v_{i\tau} = a_{1it} y_{gap,i\tau} \frac{\overline{g_{i\tau}}}{y_{i\tau}} + a_{2it} \{\ln(g_{i\tau}) - \overline{\ln(g_{i\tau})}\} \frac{\overline{g_{i\tau}}}{y_{i\tau}} + a_{3it} \frac{b_{i,\tau-1}}{y_{i\tau}} + a_{4it} + \varepsilon_{i\tau},$$

for $\tau \in (t-5, t+4)$.

that is, one uses a ten year centered rolling window to estimate the pro-cyclicality of the budgetary variable at any date t . This method suffers however from serious shortcomings. First, by definition, we lose the first five years and the last four years of data for each country. Second, because the method involves estimating a coefficient by discarding at each time period one old observation and taking into account a new one, the coefficient can vary substantially when the new observation is very different from the one it replaces. This implies that the series may be jagged and affected by noise and transitory changes; moreover, a sudden jump in the series would not be coming from changes in the immediate neighborhood of date t , but from changes 5 years before and 4 years after. To

⁶In the remainder of the paper, we will often loosely talk about procyclicality or cyclicity of public debt, investment or consumption. All these terms always refer to the a_1 coefficient in equation 1 when estimated with a left-hand side variable based on either debt, investment or consumption.

avoid these problems we decided instead to implement two alternative methods for our first stage estimations.

The first method we use is local Gaussian-weighted ordinary least squares. The method is very similar in spirit to the 10-years rolling window. But, instead of taking only ten observations at a time, we take all the observations and we weigh them by a Gaussian centered at date t . Again the regressions are run separately for each country i .

$$v_{i\tau} = a_{1it} y_{gap,i\tau} \frac{\overline{g_{i\tau}}}{y_{i\tau}} + a_{2it} \{\ln(g_{i\tau}) - \overline{\ln(g_{i\tau})}\} \frac{\overline{g_{i\tau}}}{y_{i\tau}} + a_{3it} \frac{b_{i\tau-1}}{y_{i\tau}} + a_{4it} + \varepsilon_{i\tau},$$

where $\varepsilon_{i\tau} \sim N(0, \sigma^2/w_t(\tau))$ and $w_t(\tau) = \frac{1}{\sigma\sqrt{2\Pi}} \exp\left(-\frac{(\tau-t)^2}{2\sigma^2}\right)$.

In practice, we use $\sigma = 5$. The choice was made to obtain sufficient smoothing of the estimates. While there is some arbitrariness in this choice, choosing 10 years rather than 8 or 12 is just as arbitrary in the case of the 10-years rolling window method. By contrast with the latter method, weighed regressions allow us to use all the data points for our estimates. Moreover, under this method, jumps in the coefficients are mainly due to changes in the immediate neighborhood of date t , as those observations in the immediate neighborhood of date t are given highest weight.

The second and preferred method, assumes that coefficients, instead of being independant over time as in ordinary least squares regressions, follow an AR(1) process, namely, using the notation from equation 1, for each country i and for each coefficient j :

$$a_{jit} = a_{ji,t-1} + \varepsilon_{it}^{a_j}, \varepsilon_{it}^{a_j} \sim N(0, \sigma_{a_j}^2). \quad (2)$$

The main challenge in this case is to estimate $\sigma_{a_j}^2$ (the variance of the coefficient) at the same time as the variance of the observation, i.e. the variance σ_ε^2 in the formulation of equation 1. Once these variances are estimated, applying the Kalman smoother gives the best estimates for a_{jit} . These equations can be estimated separately for each country using the VC (for varying coefficients) program by Ekkehart Schlicht⁷. This program uses a method to estimate variances that gives estimates asymptotically equivalent to maximum likelihood estimates⁸. The optimal solution is hard to compute as it involves

⁷The program and its documentation are downloadable from: http://www.semverteilung.vwl.uni-muenchen.de/mitarbeiter/frameset_es.htm.

⁸Because of the similarity to a maximum likelihood method such as the EM (Expectation-Maximization) algorithm, this program runs into the local maximum problem. That is, it is in some cases possible to maximize likelihood by assuming that the variance is zero. While this is indeed a solution to the problem, it is only a local maximum. In this paper, we are seeking estimates that show the variation in time of the policy variables of interest. This technical problem is then also a conceptual problem: we do not believe a variance of zero to be a reasonable solution, as it implies time-invariant estimates. It is possible to use an approximate method that mitigates this problem using the Schlicht program, and we have done so in a previous version of the paper.

Bayesian inference in a complicated hierarchical model. While finding analytical closed form solutions turns out to be virtually impossible, Markov Chain Monte Carlo (MCMC) methods provide a feasible numerical approximation. We implement the method in Matlab, assuming, in a panel-like fashion, that the variances of the coefficients and equations are the same for all countries. We are thus left with five variances to estimate: four for the coefficient processes ($\sigma_{a_j}^2, j = 1, 2, 3, 4$) and one for the variance of the error in the equation (σ_ε^2). Intuitively, the MCMC method explores randomly (using a Markov chain, hence the name) a wide spectrum of possible values for the variances, and one then retains a set that is representative of probable values given the data. In practice, the estimates obtained using this method are highly correlated with the modified Schlicht method used in a previous version of this paper. The advantage of the MCMC method is that it does not get stuck in local solutions and properly represents uncertainty about the variances. The results obtained from this technique⁹ thus do not suffer from the usual overfitting problems and the data analysis becomes more meaningful.

The three estimation methods for the first stage entail different statistical properties and underlying assumptions about the processes at play. From a statistical point of view, the rolling window method relies on less data than the local Gaussian-weighted OLS and AR(1) approaches. The last two approaches are more reliable because they use the whole available data to estimate the coefficients and they proceed in a more structured way.

The difference between the different estimation methods is meaningful from an economic point of view. Indeed, with the AR(1) method, it is assumed that the cyclicity of public spending only changes slowly over time and always follows the same underlying laws as defined by the estimated variances of the coefficients. By contrast, in the case of the rolling window approach, a sudden substantial change in the cyclicity of public spending is not ruled out. Moreover, no persistence of underlying parameters is assumed in the long run: indeed, the coefficient estimated, say 15 years after the start of the data series uses a completely different set of observations from the coefficient estimated 5 years after the start of the data series. The Gaussian-weighted OLS lies between the AR(1) and the 10-years rolling window methods: on the one hand, it does not directly assume any persistence of the coefficient so that the latter is allowed to change in an unconstrained way; on the other hand, the wider the Gaussian used for weighing observations, the smoother the coefficient estimates over time. Thus, "small" changes (or noise) are smoothed out, but yet "big" changes remain visible, the meaning of big and small depending on the width of the Gaussian.

Overall, we believe that the assumptions underlying the Gaussian-weighted OLS and the AR(1) model are more reasonable than the OLS rolling window both from a statistical and economic perspective, hence our choice to emphasize Gaussian-weighted OLS and AR(1) results in the text, while the 10-years rolling window results are relegated in the appendix 3. Appendix 2 shows the results

⁹See the technical appendix for more details on the implementation of this method.

from the procyclicality estimations with all three methods.

2.4 Results

We now use the Gaussian-weighted OLS and AR(1) methods to characterize the level and time path of the procyclicality of government debt in the OECD countries in our sample.

Table 1 summarizes the descriptive statistics of our main variables of interest. It is worth noting that the three different methods used in the first stage to estimate procyclicality give very similar results. We note that gross debt is countercyclical (negative coefficient) while public consumption and investment are procyclical (positive coefficient). These findings are consistent with Lane (2003), who finds that government investment and consumption are procyclical, and that the primary surplus is procyclical, which in turn is equivalent to saying that government debt is countercyclical. Moreover, the mean of our gross debt procyclicality estimate is very close to -1 for all methods considered, which is in line with the tax smoothing model described above.

TABLE 1 HERE

We now look at the evolution of the procyclicality of public debt growth, as measured by the estimated coefficients a_{lit} from equation 1. Figure 1 shows the evolution of the procyclicality of debt for the US estimated by the three methods described above. We can readily see that, as expected, the 10 years rolling window yields most volatile results, and the AR(1) method the smoothest with the Gaussian-weighted OLS method lying in between. Overall, all three methods show a decrease in procyclicality over time, with a recent trend towards increasing procyclicality shown by the 10 years rolling window and Gaussian-weighted OLS methods.

FIGURE 1 HERE

In Figure 2, we then show the procyclicality of public debt estimated through the AR(1) method for a few countries in our sample. In general, and in line with US trends, procyclicality tends to diminish over time, especially since the 1980's. This downward trend in procyclicality is however more pronounced for the UK and the US than for the average of EMU countries. Also, one can observe some divergence between EMU and non-EMU countries: at the beginning of the period, the procyclicality of public debt growth in EMU countries was very similar to that in the US or the UK, however, as of the 1990's, the US and the UK became significantly more countercyclical whereas the EMU did not.

FIGURE 2 HERE

In Figure 3, we plot the same evolution, however based on coefficients that are estimated using the Gaussian-weighted OLS. As can be seen, trends in estimates are very similar to those obtained using the AR(1) method.

FIGURE 3 HERE

These results are consistent with Gali and Perotti (2003), who show, splitting their sample by decades, that in general fiscal deficits in the OECD have become more countercyclical, but less so in EMU countries. Here, we confirm these results using a full-fledged time-series measure of cyclicality.

To summarize our results from first stage regressions, we found that government debt has become more countercyclical in non-EMU countries than in EMU countries since the 1990s.

3 Second stage regressions: the effect of countercyclical budgetary policies on growth

In this section we regress growth on the cyclical coefficients derived for each budgetary variable in the first stage regressions of the previous section, financial development, the interaction between the two variables, and a set of controls. Our conjecture is that the more firms are credit constrained, that is, the lower financial development, the more growth-enhancing countercyclical budgetary policies should be to the extent that they reduce the costs that negative liquidity shocks impose on credit-constrained firms. The underlying idea, modelled by AABM, is that, in an economy with tight credit constraints, the occurrence of a recession forces a number of firms to cut on innovative investments in order to survive idiosyncratic liquidity shocks.

To reduce the negative consequences of a recession (or a bad aggregate shock) on firms' innovative investments, the government may decide to increase the volume of its public investments, or public consumption, or to subsidize consumer credit in order to foster the demand for private firms' products. Alternatively, the government may choose to directly increase its subsidies to private enterprises, thereby increasing their liquidity holdings and thus making it easier for them to face idiosyncratic liquidity shocks without having to sacrifice R&D or other types of longer-term growth-enhancing investments. However, this may have the perverse effect of softening firms' budget constraints, thereby partly undermining the potential innovation-enhancing effect of recessions.

That government intervention might increase aggregate efficiency in an economy subject to credit constraints and aggregate shocks, has already been pointed out by Holmstrom and Tirole (1998). Our analysis in this section can be seen as a first attempt to explore potential empirical implications of this idea for the relationship between growth and public spending over the cycle.

3.1 Theoretical background

The following toy model is directly adapted from AABM and ABRR. Consider an economy composed of a continuum of firms, each of which lives for two periods. A firm born at date t produces at that date according to

$$y_t = a_t,$$

where a_t denotes the knowledge adjusted level of aggregate productivity. At the beginning of date t , the firm can also invest in R&D. Investing R&D effort $\frac{1}{2}z^2$, allows the firm innovate in period $(t + 1)$ with probability z , provided the firm overcomes an idiosyncratic liquidity shock occurring at the end of period t . For simplicity, suppose that the liquidity shock \tilde{c} is independently and identically distributed across firms with uniform distribution over the interval $[0,1]$, whereas the aggregate shock a_t over time is distributed according to

$$a_t = \begin{cases} 1 + \varepsilon & \text{with probability } 1/2 \\ 1 - \varepsilon & \text{with probability } 1/2 \end{cases}.$$

The long-term R&D investment yields a (knowledge-adjusted) value equal to $\nu > 0$ in period $(t + 1)$ whenever innovation is successful. The investment decision is made before the realization of the aggregate shock a_t . Finally, credit market imperfections prevent a firm with short-run profit flow a to invest more than μa , where $1 < \mu < \infty$, for the purpose of covering its idiosyncratic liquidity cost \tilde{c} .

Before aggregate productivity a_t is realized, firms will choose to invest in R&D the amount of effort

$$z^* = \arg \max \left\{ Vz - \frac{1}{2}z^2 \right\} = V,$$

where

$$V = \nu \cdot \mathbb{E}_t(\min[1, \mu a_t]),$$

with \mathbb{E}_t denoting to the expected value at date t , and where

$$\min[1, \mu a_t] = \Pr(\tilde{c} \leq \mu a_t)$$

is the probability of the firm overcoming its liquidity shock in period t conditional upon a_t .

One can easily show that a mean-preserving spread of a_t will reduce V and therefore the firm's incentive to invest in R&D as it will reduce the expected probability of overcoming the liquidity shock. It will thus also reduce more the expected growth rate which we take to be equal to the expected innovation flow:

$$g_t = z^* \mathbb{E}_t(\min[1, \mu a_t]) = \nu \{ \mathbb{E}_t(\min[1, \mu a_t]) \}^2.$$

A countercyclical public debt policy that consists in taxing individuals when $a_t = 1 + \varepsilon$ in order reduce the incidence of a low $a_t = 1 - \varepsilon$ on firms' short term profits, should then be growth-enhancing, and all the more so when μ is lower.

3.2 Empirical specifications

In all the specifications we use for our second-stage regressions, we measure productivity growth by the first difference of the log of real GDP per capita. We shall regress this left hand side variable on the lagged cyclicity of public debt growth as derived in the first stage regressions, the lagged private credit

measure captured by the ratio of private credit to GDP and borrowed from Levine (2001), and the interaction between those two variables. As control variables, we use the lag of log real GDP per capita, the level of schooling, openness to trade, inflation¹⁰, population growth, and the government share of total GDP. Moreover, in all specifications and unless otherwise specified, we weigh each observation by the inverse of the variance of the estimated cyclicity coefficient (aweights in Stata), thus giving higher weight to coefficients that are more precisely estimated in the first stage.

Each of the tables we present here are structured as follows. Using the set of cyclicity measures derived in the first stage respectively from the Gaussian-weighted least squares and AR(1) methods, we first perform ordinary least squares regressions (first column). We then move on to country fixed effect estimates (second column). As is typical in panel growth regressions and can be confirmed by the Wooldridge test implemented in Stata's xtserial command, the errors are serially correlated (AR(1)) in first differences¹¹. This implies that country fixed effect estimates may be biased. To correct for this potential source of bias, we use Stata's xtregar command, which implements the method described in Batalgi (2001) to estimate the coefficient of correlation between the errors and give unbiased estimates (third column). Xtregar allows the use of weights¹², but the weights can only be country specific and not observation specific¹³. We therefore weigh observations by the inverse of the average variance of first-stage estimates for each country. Lastly, we include year fixed effects on top of country fixed effects and estimate these specifications, respectively without and with correction for AR(1) in the error term (fourth and fifth columns).

3.3 Results

Table 2 shows the results of regressing the first difference of the log of real GDP per capita over the lagged cyclicity of government debt, measured by the coefficients obtained in the first stage regression using the AR(1) method. The prediction is that of a negative coefficient for the effect on growth of the procyclicality of public debt, and of a positive coefficient on procyclicality interacted with financial development, and we see that the corresponding coefficients in Table 2 always have the desired signs.

However, it is only when we control for country or country year fixed effects that the results become significant and they are indeed significant at the 5% or

¹⁰While we do not directly address monetary policy in this work, controlling for inflation allows to indirectly take into account the effect of monetary policy on growth.

¹¹This is true whether or not we include the lagged dependant variable, i.e. lag of log real GDP per capita.

¹²In principle, GMM Arellano-Bond or Blundell-Bond estimates could also usefully address this problem. In practice, these methods as implemented with Stata do not allow for weights, which makes them considerably less useful in the context of this paper. We did however try to use these methods without weights, but they never passed essential specification tests, which is one further reason not to rely on them in this specific case.

¹³Moreover, the use of weights restricts the range of methods available to calculate the coefficient of the AR(1) process in the errors. We chose the "regress" method here.

even at the 1% level. The magnitude of coefficients also increases when adding country fixed effects as compared to OLS. In the last two columns, the further addition of year effects slightly decreases the magnitude of the coefficients and their significance. Allowing for an AR(1) structure in the error term in columns 3 and 5 does not significantly affect the results.

Table 2 is thus consistent with the prediction of a negative effect of procyclicality in public debt on growth, whereas we see a positive and significant interaction effect between private credit and the procyclicality variable. Thus the less financially developed a country is, the more growth-enhancing it is for the government to be countercyclical in its debt policy.

TABLE 2 HERE

Table 3 below repeats the same exercises as in Table 2 but now the lagged cyclicity of public debt measures are those derived from first-stage regressions using the Gaussian-weighted least squares method. Results are very similar in magnitude and significance to the ones presented in Table 2. The only noticeable difference is that the coefficient on the interaction term between procyclicality of public debt and private credit over GDP is smaller and no longer significant when adding both country and year fixed effects. This might be due to the Gaussian weighted method being noisier than the AR(1) method, a presumption which is reinforced by the fact that moving from the Gaussian to the even noisier 10-year OLS rolling window method removes the significance of all coefficients, as shown in Table 1 in the Appendix 3. However, it must be noted that part of this loss of significance is due to a higher sensitivity of results to the inclusion of the lagged dependant variable (that is $\text{lag}(\log(\text{real GDP per capita}))$) in the case of the 10-years rolling window method. This arises because there non negligible correlation between lagged private credit over GDP, lagged log of GDP per capita and the cyclicity measure, implying that including the lagged dependant variable worsens multicollinearity problems. In Table 2 of Appendix 3, we thus repeat the same exercise as in Table 1 of Appendix 3, and find that results are more often significant when excluding the lagged dependant variable. Results are now significant for both the main term and the interaction with country fixed effects, and remain significant for the interaction when including year fixed effects while the main effect is then only significant at 22%. It is still the case however that the significance disappears when accounting for the AR(1) structure in the error term in columns 3 and 5. Even with the 10 years rolling window method, a cruder and noisier estimation procedure, we thus find some evidence that less countercyclical public debt policy is harmful for economic growth, and even more so when financial development is lower.

TABLE 3 HERE

To get a better sense of the magnitude of the effects of public debt cyclicity on growth and the interaction of these effects with financial development, we can ask the following question: according to our estimates, what would happen if public debt in the EMU became as countercyclical as that in the US? Table

4 summarizes the answer to this question, which is based on the estimates in columns 5 of Tables 2 and 3, that is on the most demanding specification with both country and year fixed effects and correction for an AR(1) structure in the error term. Thus, if the EMU's government debt was to become as countercyclical as that in the US, which corresponds to a reduction in procyclicality equal to 1.18 units, then the EMU would gain 0.72 points of growth if using the AR(1) method (Panel A) and 0.84 points of growth when using the Gaussian-weighted OLS method (Panel B). By contrast, if the US was to reduce its procyclicality by the same 1.18 units, it would grow *less* by up to 5.7 points (Panel A).

TABLE 4 HERE

Next, we investigate whether these results are driven more by government investments, or by government consumption. This part of our analysis is highly tentative and exploratory, and in particular it relies on the ad hoc assumption that the cyclicity of the various components of public spending would respond to the same right hand side variables as the growth of public debt and according to the same kind of equation. Tables 5 and 6 show the results of the analysis for respectively public consumption and investment using the AR(1) MCMC method. Results are broadly insignificant for public consumption (Table 5), while for public investment (Table 6) they are similar to those found for public debt. However, a procyclical public investment seems to be significantly diminishing growth irrespective of the level of financial development, as the interaction between the procyclicality of public investment and private credit over GDP is typically insignificant. On the other hand, when using the Gaussian-weighted rolling window for the second stage (Tables 3 and 4 in Appendix 3) we find the results for public consumption to be more similar to those for public debt than the results for public investment. We therefore conclude that there is some evidence that procyclical public investment and consumption are harmful for growth, and possibly even more so when a country is less financially developed. However, these results are more tentative and sensitive to the first-stage methods and specifications used than in the case of public debt. Part of the reason for less clear-cut results might be stemming from the fact that it is inadequate to use the same first-stage specification for public consumption and investment as for public debt.

TABLE 5, 6 HERE

One interesting conclusion from the above analysis, is that EMU countries should have public debt policies that are more countercyclical, or at least as countercyclical, than in the US. On the other hand, the US are more financially developed than the EU (the ratio of private credit to GDP in 2000 in the EU is equal 0.92 against 2.17 in the US, and this difference abstracts from differences in stock market and venture capital market development), and therefore one might wonder whether growth in the US might actually benefit from a reduction in the countercyclicality of public debt policy. Our regressions in Tables 2 and 3 and our calculations in Table 4 suggest that this might indeed be the case: since

the US decrease growth by getting more countercyclical, the linear specification used in our model implies that at their level of financial development they could increase growth by becoming less countercyclical.

Finally, these conclusions raise the issue of whether or to which extent individual countries can control the degree of public debt countercyclicality. Part of the answer may involve the cost at which governments can borrow from domestic or foreign capital markets, and whether this cost is itself procyclical or countercyclical. This brings us to the following section on the determinants of the cyclicity of public debt.

4 Determinants of the cyclicity of public debt

We now investigate possible explanations for the observed differences in the procyclicality of public debt policy across countries in our sample and over time. Since our sample is restricted to OECD countries, little variation should be expected from the corruption or other institutional variables considered by the literature so far¹⁴. Instead, we will focus on potential explanatory variables such as financial development or openness, that may affect governments' ability to borrow over the cycle or the cost for them to do so. We also include GDP growth volatility as measured by the standard error of GDP growth, lag of log real GDP per capita, the government share of GDP, and EMU membership¹⁵ as control variables. Financial development is a plausible suspect as it influences both the ability and the willingness of governments to borrow. While OECD countries are arguably less subject to borrowing constraints than other countries in the world, there is still a fair amount of cross-country variation in financial development among OECD countries. Openness is also a plausible candidate as one can expect a higher capital inflow in more open economies during expansion periods, and therefore a lower cost of capital during such periods. This in turn tends to increase the long-run cost of financing countercyclical public debt policies while maintaining the overall debt constant on average over the long run. The EMU dummy is also a plausible candidate, given: (i) our casual observation in the introduction to the effect that the structural deficit appears to be less reactive to the business cycle in the eurozone than in the US or the UK; (ii) the deficit and debt restrictions imposed by the Stability and Growth Pact and also the restrictions that individual countries imposed themselves in order to qualify for EMU membership.

Table 7, where the cyclicity measures are derived using the AR(1) and Gaussian-weighted methods, shows results that are consistent with these conjectures, namely: (i) higher financial development is positively and significantly correlated with the countercyclicality of government debt (the table shows a neg-

¹⁴As mentioned above, using ICRG indicators turns out not to be of interest for our analysis.

¹⁵This dummy variable takes a value of 1 for all countries that currently belong to the EMU, and 0 for all the other countries. This is because the EMU has been prepared for many years so that the countries that would eventually join might be different even before the EMU is fully effective.

ative coefficient of public debt procyclicality); in other words, it is precisely when the countercyclicality of public debt is more positively correlated with growth, namely when financial development is low, that public debt countercyclicality seems hardest to achieve; (ii) more trade openness is negatively and significantly correlated with public debt countercyclicality (the table shows a positive correlation between openness and public debt procyclicality); (iii) EMU countries appear to have a harder time achieving public debt countercyclicality. The effect of the EMU dummy is more likely to be explained by rigidities already imposed by the precursor EMS regime and then reinforced by the Maastricht Treaty, rather than the 1999 implementation of the EMU itself¹⁶; further investigation of this question is however beyond the scope of this paper. Finally, we note that higher volatility of GDP growth and a higher share of government in the GDP are associated with a more countercyclical debt policy.

TABLE 7 HERE

To summarize our discussion in this section, a lower level of financial development, a higher degree of openness, and belonging to the EMU group, enter significantly as explanatory variables to explain a lower degree of countercyclicality in government debt.

5 Conclusion

In this paper we have analyzed the dynamics and determinants of the cyclicity of public debt on a yearly panel of OECD countries, and we have analyzed the relationship between public debt countercyclicality, financial development, and productivity growth. Our findings can be summarized as follows: first, countercyclicality has increased over time across all countries in our sample, however to a lower extent in EMU countries than in the US or the UK. Second, countercyclicality of government debt appears to be facilitated by a higher level of financial development while it appears to be complicated by a higher degree of openness to trade. Third, our main finding is that countercyclical public debt policy is more growth enhancing the lower the country's level of financial development.

Our analysis may be criticized on several grounds. First, one might dispute the causal link from public debt countercyclicality to growth, arguing that our second stage regressions simply reflect the fact that growth allows a country to become more countercyclical. Without ruling out the existence of a reverse causality from growth to public debt policy over the cycle, and in the absence of a good experiment or instrument, the case can still be made to support the claim that reverse causality is not plaguing our main results. First, if all there is, is that countries with higher average growth can afford to be more countercyclical, then country fixed effects would account for this. Second, we use lagged and

¹⁶We have thus experimented with an interaction between the EMU dummy and a post-1999 dummy, but this interaction was typically insignificant, indicating that there is no substantial change occurring with the full implementation of the EMU in 1999.

not current policy cyclical; in fact, we can show that future cyclical is not significantly correlated with growth whereas past cyclical is so correlated at several lags, thereby further supporting the idea that cyclical Granger-causes growth and not the reverse.

More generally, we have an endogeneity problem if unobserved determinants of cyclical or financial development also influence growth. If these determinants are country-specific and time-invariant, once again the inclusion of country fixed-effects cure it. Moreover, year fixed effect account for unobserved determinants that are time-varying and common to all countries. Thus, the endogeneity problem only arises with time-variant country-specific factors that are unobserved or uncontrolled for. What our analysis suggests is that openness cannot be such a factor, as higher openness is negatively correlated with public debt countercyclical whereas it is positively correlated with growth.

Another potential problem arises from the fact that because GDP growth and GDP gap are positively correlated, the effect of procyclical (estimated coefficient on GDP gap) on growth will mechanically tend to be negative, creating a spurious negative correlation between procyclical and growth automatically by construction. However, we use the lagged value of procyclical and, while current growth and GDP gap are indeed positively correlated, current growth and the lagged GDP gap are uncorrelated¹⁷.

Finally, in light of our second stage regressions, one may wonder whether the EMU should really change its policy given that financial development is increasing in the EMU over time, boosted by the monetary union itself. However, based on the historical growth trend of financial development, the EMU will not reach the current level of financial development of the USA until year 2084. Even based on the faster growth trend of financial development of the 1990's, the EMU will only reach the USA level of financial development in 2068. On the other hand the USA, according to our estimates, might be too countercyclical. In order for more countercyclical to be no longer growth enhancing for the EMU, and based on growth trend of financial development of the 1990's, the EMU would still have to wait for several decades (at least until between 2020 and 2030, depending on which specification we base our estimates on).

We now conclude by mentioning some possible research avenues. First, one could try to perform the same kind of analysis for other groups of countries, e.g middle income countries in Latin America or in Central and Eastern Eu-

¹⁷While this makes us fairly confident in our results, it does not completely rule out some degree of spurious correlation between GDP growth and procyclical. Indeed, by construction our first stage estimates of procyclical do not rely only on the current value of the right-hand side variables (including GDP gap). In the Gaussian weighted OLS method, current GDP gap is however given more weight than future or past GDP gaps, so lagged procyclical depends more on lagged than current GDP gap. In the AR(1) method, by the AR(1) assumption the value of the current cyclical coefficient depends only on the past value of the cyclical coefficient and not on future values. Thus lagged cyclical depends primarily on lagged GDP gap and even further lags of the GDP gap, and not on the current GDP gap. The use of the Kalman smoother does however introduce some small degree of feedback between future and past values of the coefficient. However, all in all, the AR(1) method should be resistant to the bias arising from spurious correlation.

rope. Second, one could explore potential implications of the relationship between growth and the countercyclicality of budgetary policy for the conduct of monetary policy. For example, to which extent allowing for some inflation or for higher procyclicality of short term interest rates, increases governments' borrowing costs during recessions, and thereby improves their ability to implement growth-enhancing countercyclical budgetary policies? Preliminary work with Enisse Kharroubi suggests that inflation is less negatively correlated with growth, the higher a country's degree of public debt procyclicality. Third, one could analyze in more detail which types of countercyclical public spending (consumption, investment) are most growth-enhancing, and on which sectors. Finally, one could investigate the possible interactions between countercyclical budgetary policy and structural reforms in the product and labor markets.

References

- [1] Acemoglu, D, Johnson, S, Robinson, J, and Y. Thaicharoen (2003), "Institutional Causes, Macroeconomic Symptoms: Volatility, Crises, and Growth", *Journal of Monetary Economics*, 50,49-123.
- [2] Aghion, P, Angeletos, M, Banerjee, A, and K. Manova (2005), "Volatility and Growth: Credit Constraints and Productivity-Enhancing Investment", NBER Working Paper No 11349.
- [3] Aghion, P, Bacchetta, P, Ranciere, R, and K. Rogoff (2006), "Exchange Rate Volatility and Productivity Growth: The Role of Financial Development", NBER Working Paper No 12117.
- [4] Alesina, A, and R. Perotti. (1996), "Fiscal Adjustments in OECD Countries: Composition and Macroeconomic Effects", NBER Working Paper No 5730.
- [5] Alesina, A, and G. Tabellini (2005), "Why is Fiscal Policy often Procyclical?", NBER Working Paper No 11600.
- [6] Barro, R (1979), "On the Determination of Public Debt", *Journal of Political Economy*, 87, pp. 940-971.
- [7] Baltagi, B (2001), *Econometric Analysis of Panel Data*, 2d ed. New York: John Wiley & Sons.
- [8] Calderon, C, Duncan, R, and K. Schmidt-Hebbel (2004), "Institutions and Cyclical Properties of Macroeconomic Policies", Central Bank of Chile Working Papers No 285.
- [9] Chib,S, and E. Greenberg (1995), "Understanding the Metropolis-Hastings Algorithm", *The American Statistician*, 49, 327-335.

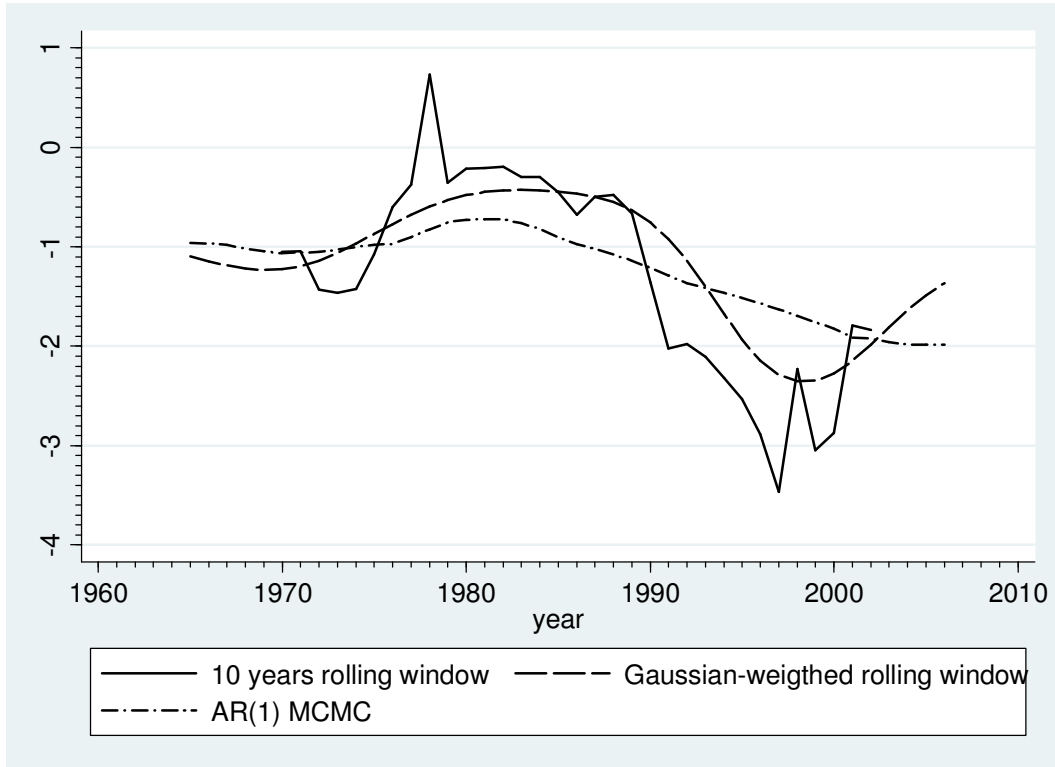
- [10] Easterly, W, (2005), "National Policies and Economic Growth: A Reappraisal", Chapter 15 in *Handbook of Economic Growth*, P. Aghion and S. Durlauf eds.
- [11] Gali, J, and R. Perotti (2003), "Fiscal Policy and Monetary Integration in Europe", *Economic Policy*, 533-572.
- [12] Kording K. and Marinescu I. (2006), "Sampling approaches to Kalman learning", work in progress.
- [13] Lane, P, (2002), "The Cyclical Behavior of Fiscal Policy: Evidence from the OECD", *Journal of Public Economics*, 87, 2661-2675.
- [14] Lane, P, and A. Tornell (1998), "Why Aren't Latin American Savings Rates Procyclical?", *Journal of Development Economics*, 57, 185-200.
- [15] OECD (1995), "Estimating Potential Output, Output Gaps and Structural Budget Balances", by Claude Giorno, Pete Richardson, Deborah Roseveare and Paul van den Noord, OECD Economics Department Working Paper No. 152.
- [16] Talvi, E, and C. Vegh (2000), "Tax Base Variability and Procyclical Fiscal Policy", NBER Working Paper No 7499.

Table 1: Summary statistics

	Obs.	Mean	Std. Dev.	Min	Max
GDP gap	764	-0.001	0.028	-0.109	0.160
Gross government debt/GDP	756	0.501	0.283	0.044	1.608
(d.Gross government debt-interests)/GDP	754	0.008	0.041	-0.113	0.195
Government consumption/GDP	1029	0.182	0.045	0.057	0.301
d.Government consumption/GDP	1003	0.017	0.011	-0.009	0.081
Government investment/GDP	966	0.036	0.011	0.012	0.115
d.Government investment/GDP	939	0.003	0.005	-0.016	0.041
Procyclicality of gross government debt (AR(1))	612	-0.925	1.042	-4.206	1.447
Procyclicality of gross government debt (10-years rolling window)	454	-1.179	1.923	-9.259	3.466
Procyclicality of gross government debt (weighted rolling window)	643	-1.072	1.169	-4.000	1.951
Procyclicality of government consumption (AR(1))	605	0.126	0.209	-0.850	0.771
Procyclicality of government consumption (10-years rolling window)	425	0.133	0.240	-0.767	0.891
Procyclicality of government consumption (weighted rolling window)	605	0.143	0.161	-0.368	0.584
Procyclicality of government investment (AR(1))	605	0.073	0.061	-0.108	0.286
Procyclicality of government investment (10-years rolling window)	425	0.093	0.142	-0.588	0.574
Procyclicality of government investment (weighted rolling window)	605	0.077	0.092	-0.314	0.399
Growth of GDP per capita	1117	0.027	0.029	-0.942	0.199
Private credit/GDP	983	0.679	0.437	0.013	2.370
Average years of schooling for the population over 25 years old	1078	7.691	2.285	1.940	12.250
Openness	1242	50.131	40.938	3.648	266.883
Inflation	1212	0.085	0.125	-0.105	1.408
Population growth	1132	0.008	0.008	-0.038	0.047
Government share of GDP (in %)	1241	13.292	5.779	3.008	32.115
Investment/GDP (in%)	1241	23.325	5.407	8.208	41.635

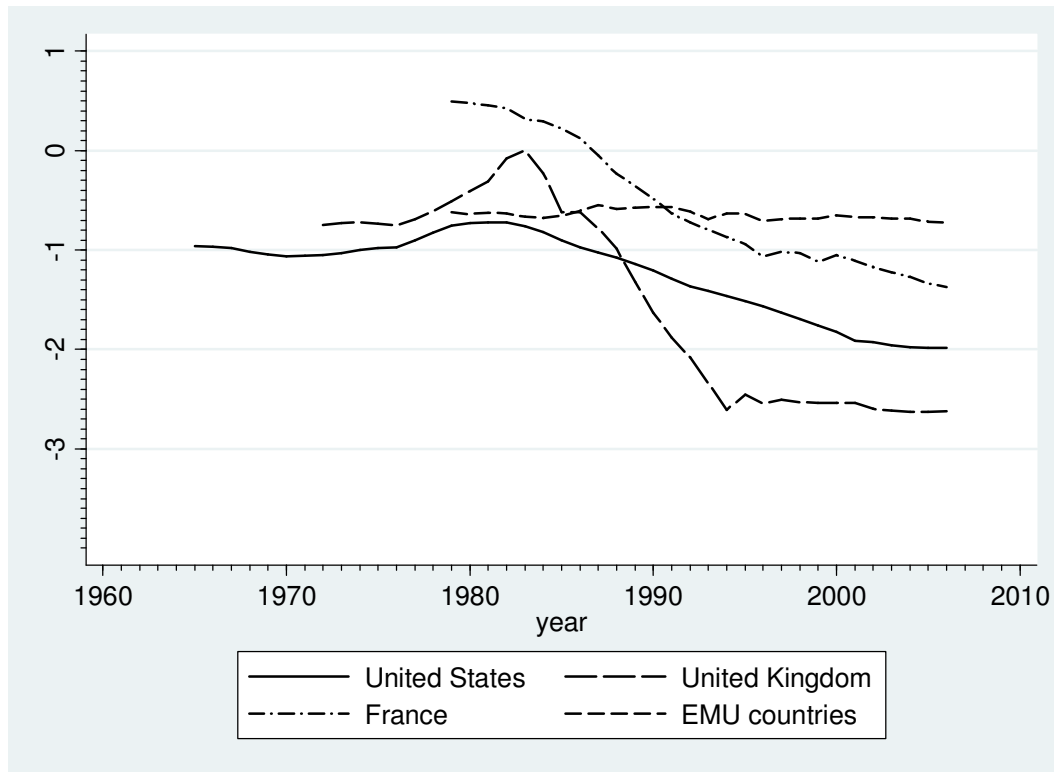
Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Figure 1: the procyclicality of public debt in the USA



Note: the graph plots the a_{lit} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques.
Source: OECD Economic Outlook.

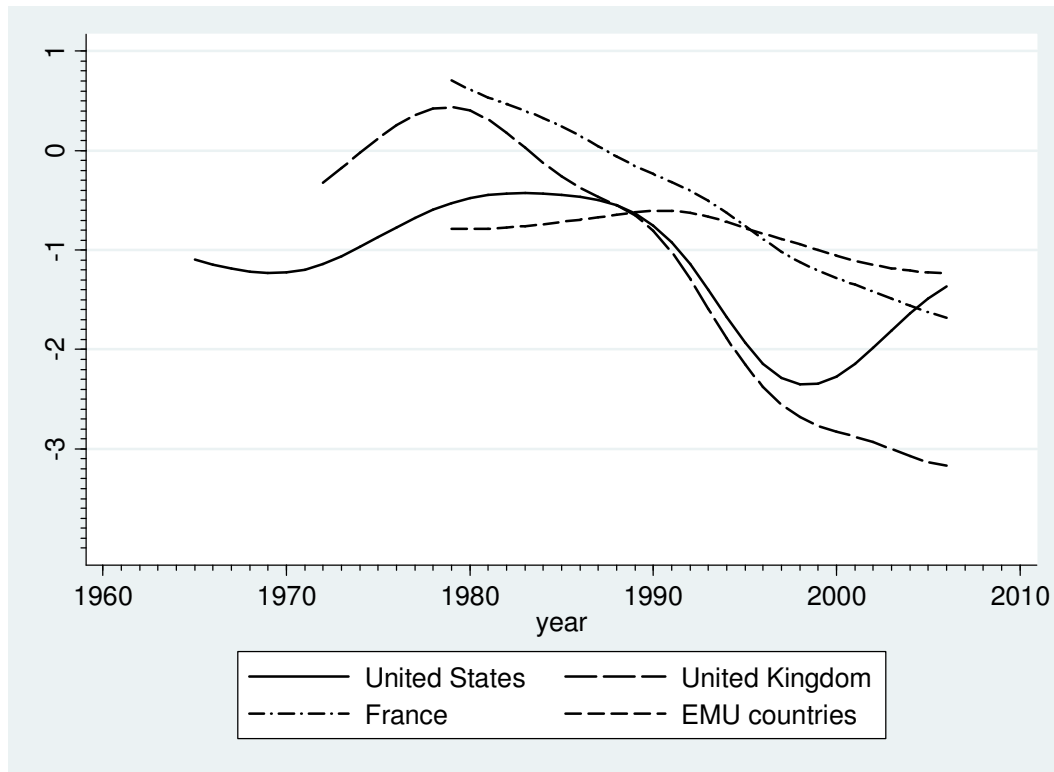
Figure 2: The procyclicality of public debt using the AR(1) MCMC method



Note: the graph plots the a_{1it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using the AR(1) MCMC method. For EMU countries (i.e. countries who are or will be part of the EMU), the line represents the average of the estimated coefficients for the EMU countries present in the sample; the average is only computed for those years where all EMU countries have non-missing observations.

Source: OECD Economic Outlook.

Figure 3: The procyclicality of public debt using the Gaussian-weighted OLS method



Note: the graph plots the a_{1it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using the Gaussian-weighted rolling window OLS method. For EMU countries (i.e. countries who are or will be part of the EMU), the line represents the average of the estimated coefficients for the EMU countries present in the sample; the average is only computed for those years where all EMU countries have non-missing observations.

Source: OECD Economic Outlook.

Table 2: The effect of public debt procyclicality on growth, AR(1) MCMC method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government debt)	-0.004 (0.002)*	-0.023 (0.005)***	-0.022 (0.008)***	-0.015 (0.005)***	-0.018 (0.007)**
lag(Private credit/GDP)	0.000 (0.005)	-0.003 (0.009)	0.010 (0.012)	-0.012 (0.009)	-0.011 (0.011)
lag(Procyclicality of government debt*Private credit/GDP)	0.004 (0.003)	0.017 (0.005)***	0.017 (0.007)**	0.011 (0.005)**	0.014 (0.007)**
lag(log (real GDP per capita))	-0.001 (0.001)	-0.098 (0.015)***	-0.177 (0.020)***	-0.134 (0.023)***	-0.276 (0.030)***
Average years of schooling for the population over 25 years old	-0.001 (0.001)	0.007 (0.003)**	0.018 (0.004)***	0.005 (0.003)**	0.009 (0.004)**
Openness	0.000 (0.000)	0.001 (0.000)***	0.001 (0.000)***	0.001 (0.000)**	0.000 (0.000)
Inflation	-0.047 (0.021)**	-0.071 (0.021)***	-0.094 (0.022)***	-0.043 (0.022)*	-0.055 (0.021)**
Population growth	-1.250 (0.405)***	-2.085 (0.317)***	-2.278 (0.246)***	-1.481 (0.283)***	-1.571 (0.230)***
Government share of GDP (in %)	0.000 (0.000)	-0.001 (0.000)**	-0.001 (0.000)**	-0.001 (0.000)*	-0.001 (0.000)*
Investment/GDP (in%)	0.001 (0.000)***	0.003 (0.000)***	0.004 (0.000)***	0.002 (0.000)***	0.004 (0.000)***
Constant	-0.005 (0.021)	-0.906 (0.136)***	-1.672 (0.093)***	-1.151 (0.192)***	-2.242 (0.113)***
Observations	460	441	441	441	441
R-squared	0.17	0.4		0.61	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 3: The effect of public debt procyclicality on growth, Gaussian-weighted OLS method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government debt)	-0.003 (0.003)	-0.029 (0.005)***	-0.023 (0.009)***	-0.010 (0.004)**	-0.009 (0.010)
lag(Private credit/GDP)	-0.009 (0.006)	0.000 (0.010)	0.008 (0.014)	-0.026 (0.009)***	-0.031 (0.016)**
lag(Procyclicality of government debt*Private credit/GDP)	-0.001 (0.004)	0.020 (0.005)***	0.015 (0.008)*	0.003 (0.004)	0.003 (0.008)
lag(log (real GDP per capita))	-0.001 (0.002)	-0.108 (0.018)***	-0.264 (0.023)***	-0.149 (0.028)***	-0.508 (0.037)***
Average years of schooling for the population over 25 years old	0.001 (0.001)	0.004 (0.005)	0.021 (0.005)***	0.001 (0.004)	0.001 (0.006)
Openness	0.000 (0.000)**	0.001 (0.000)***	0.002 (0.000)***	0.001 (0.000)**	0.001 (0.000)**
Inflation	-0.048 (0.025)*	-0.077 (0.024)***	-0.084 (0.021)***	-0.032 (0.033)	-0.043 (0.020)**
Population growth	-0.492 (0.374)	-1.987 (0.500)***	-2.016 (0.256)***	-1.026 (0.375)***	-1.188 (0.225)***
Government share of GDP (in %)	0.000 (0.000)	-0.001 (0.000)**	-0.001 (0.000)***	-0.001 (0.000)**	-0.001 (0.000)*
Investment/GDP (in%)	0.002 (0.000)***	0.004 (0.001)***	0.005 (0.000)***	0.003 (0.001)***	0.006 (0.001)***
Constant	-0.022 (0.028)	-1.045 (0.187)***	-2.533 (0.091)***	-1.330 (0.235)***	-4.172 (0.079)***
Observations	485	465	465	465	465
R-squared	0.11	0.37		0.64	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 4: Implied effect on growth of a change in the procyclicality of public debt such that the EMU would have the same level of procyclicality as the US in 2000

	Estimated coef. on lag (Procyclicality of government debt)		Difference US- EMU		Estimated coef. lag(Procyclicality of government debt*private credit/GDP)		Difference US- EMU		Average(lag (private credit/ GDP))		Implied effect on growth
Panel A: AR(1)											
EMU in 2000	-0.0180	*	-1.1840	+	0.0140	*	-1.0943	*	0.9242	=	0.0072
US in 2000	-0.0180	*	-1.1840	+	0.0140	*	-2.5688	*	2.1696	=	-0.0567
Panel B: WRW											
EMU in 2000	-0.0090	*	-1.2990	+	0.0030	*	-1.2006	*	0.9242	=	0.0084
US in 2000	-0.0090	*	-1.2990	+	0.0030	*	-2.8184	*	2.1696	=	-0.0067

Note: The estimated coefficients are taken from estimates columns 5 of Tables 2 (Panel A) and 3 (Panel B). The difference US-EMU is defined as: (procyclicality of government debt in the US in 2000)- (procyclicality of government debt in the EMU in 2000). The averages of variables are calculated using the same sample on which regressions were estimated.

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 5: The effects of government consumption procyclicality on growth, AR(1) MCMC method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government consumption)	-0.003 (0.014)	0.002 (0.018)	-0.002 (0.024)	0.005 (0.014)	-0.007 (0.024)
lag(Private credit/GDP)	-0.002 (0.004)	-0.011 (0.007)	-0.007 (0.009)	-0.016 (0.006)**	-0.023 (0.010)**
lag(Procyclicality of government consumption*Private credit/GDP)	-0.021 (0.021)	-0.030 (0.024)	-0.020 (0.032)	-0.022 (0.018)	-0.007 (0.032)
lag(log (real GDP per capita))	-0.002 (0.001)	-0.086 (0.016)***	-0.147 (0.020)***	-0.122 (0.024)***	-0.274 (0.032)***
Average years of schooling for the population over 25 years old	-0.001 (0.001)	0.008 (0.003)***	0.017 (0.004)***	0.006 (0.003)**	0.010 (0.004)***
Openness	0.000 (0.000)	0.001 (0.000)***	0.001 (0.000)***	0.001 (0.000)***	0.000 (0.000)
Inflation	-0.062 (0.020)***	-0.075 (0.022)***	-0.101 (0.023)***	-0.037 (0.024)	-0.059 (0.022)***
Population growth	-0.614 (0.315)*	-1.637 (0.542)***	-1.379 (0.437)***	-1.174 (0.462)**	-0.744 (0.404)*
Government share of GDP (in %)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)
Investment/GDP (in%)	0.001 (0.000)***	0.002 (0.000)***	0.004 (0.000)***	0.002 (0.000)***	0.004 (0.001)***
Constant	-0.002 (0.020)	-0.810 (0.147)***	-1.391 (0.098)***	-1.085 (0.197)***	-2.210 (0.115)***
Observations	453	434	434	434	434
R-squared	0.13	0.32		0.56	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 6: The effects of public investment procyclicality on growth, AR(1) MCMC method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government investment)	-0.010 (0.053)	-0.336 (0.107)***	-0.398 (0.159)**	-0.324 (0.087)***	-0.464 (0.151)***
lag(Private credit/GDP)	-0.007 (0.005)	-0.022 (0.008)***	-0.008 (0.012)	-0.027 (0.007)***	-0.027 (0.012)**
lag(Procyclicality of government investment*Private credit/GDP)	0.023 (0.064)	0.069 (0.066)	-0.022 (0.114)	0.091 (0.051)*	0.018 (0.107)
lag(log (real GDP per capita))	-0.001 (0.001)	-0.092 (0.014)***	-0.168 (0.019)***	-0.132 (0.023)***	-0.295 (0.031)***
Average years of schooling for the population over 25 years old	-0.001 (0.001)	0.009 (0.003)***	0.019 (0.004)***	0.007 (0.002)***	0.011 (0.004)***
Openness	0.000 (0.000)	0.001 (0.000)***	0.001 (0.000)***	0.001 (0.000)***	0.000 (0.000)*
Inflation	-0.055 (0.020)***	-0.069 (0.020)***	-0.100 (0.022)***	-0.038 (0.023)*	-0.054 (0.021)**
Population growth	-0.662 (0.321)**	-1.393 (0.529)***	-1.206 (0.446)***	-0.936 (0.435)**	-0.646 (0.404)
Government share of GDP (in %)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)	0.000 (0.000)
Investment/GDP (in%)	0.001 (0.000)***	0.002 (0.000)***	0.004 (0.000)***	0.002 (0.000)***	0.004 (0.001)***
Constant	0.003 (0.021)	-0.838 (0.131)***	-1.561 (0.092)***	-1.143 (0.191)***	-2.374 (0.111)***
Observations	453	434	434	434	434
R-squared	0.11	0.33		0.56	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 7: The determinants of government debt procyclicality

	AR(1)			WRW		
	OLS	Country f.e.	Country year f.e.	OLS	Country f.e.	Country year f.e.
Private credit/GDP	-0.567 (0.162)***	-1.079 (0.124)***	-0.959 (0.132)***	-0.569 (0.114)***	-0.894 (0.134)***	-0.912 (0.143)***
EMU country	0.395 (0.103)***			0.165 (0.083)**		
Standard error of GDP growth	-4.929 (1.571)***			-8.684 (1.549)***		
Lag(log (real GDP per capita))	0.017 (0.040)	-0.543 (0.247)**	0.270 (0.553)	0.030 (0.048)	0.206 (0.266)	0.475 (0.502)
Openness	0.005 (0.002)***	0.010 (0.003)***	0.021 (0.005)***	-0.003 (0.001)**	-0.004 (0.004)	0.015 (0.005)***
Government share of GDP (in %)	-0.019 (0.010)**	-0.012 (0.005)**	-0.025 (0.005)***	0.004 (0.007)	-0.009 (0.006)	-0.022 (0.007)***
Constant	-0.286 (0.559)	-4.700 (2.083)**	1.443 (4.736)	0.455 (0.545)	2.123 (2.341)	3.611 (4.307)
Observations	489	489	489	515	515	515
R-squared	0.11	0.86	0.87	0.18	0.77	0.79

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the coefficient on the GDP gap composite variable from equation 1, estimated using the AR(1) MCMC method for columns 1-3, and the Gaussian-weighted rolling window method for columns 4-6. EMU country is a dummy variable equal to 1 for all countries that are part of the EMU as of 2006.

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Appendix 1: the AR(1) MCMC method for calculating cyclicity in the first stage

The aim of this section is to give a brief description of how we used the Kalman smoother together with Markov Chain Monte Carlo methods (MCMC) in order to estimate the coefficients a_{jit} from equation 1 under the assumption that they follow an AR(1) process as described by equation 2. The implementation was carried out in Matlab.

Estimating the means and variances of the coefficients of interest - that is a_{jit} in equation 2 - involves two procedures: Kalman smoothing¹ and MCMC.

To compute the coefficients with the Kalman smoother for each country, we need to know the values of five variances :

- $\sigma_{a_j}^2$ in equation 2, for $j = 1, 2, 3, 4$, i.e. the process variances in the terminology of the Kalman smoother
- the variance σ_ε^2 of the error term ε_t in equation 1, i.e. the measurement error variance in the terminology of the Kalman smoother.

Moreover, to use the Kalman smoother, we need a prior for the first period of observation for each country, that is a specification of our expectation over the values a_{jit} at the first time step. As we do not have any meaningful prior information about cyclicity at the first observed period, we use a very high variance around the prior mean, so that this prior has a negligible effect on the estimates. Specifically, the set of initial values for the coefficients were chosen to be the OLS estimates of the coefficients using the first 10 years of data for each country, and the value of the initial variance is set to be 100000 times the estimated variance of these coefficients.

However, the process variances $\sigma_{a_j}^2$ and the measurement error variance σ_ε^2 are unknown and we do not have any meaningful prior over them. We therefore need a method to find reasonable values for these five unknown variances. This is where MCMC methods are useful .

One can think of MCMC as the opposite of simulating. In the case of simulation we know the parameters of our process, for example the variances, and every time we run a simulation program, it gives us a set of possible observed data. More specifically, the probability of getting any set of observed data is the probability defined by the model that we have and the parameters. MCMC is the opposite: we assume that we have a given dataset, and we are producing a

¹For an excellent overview of the Kalman filter and smoother, see the notes by Max Welling "Kalman Filters", available on the web at <http://www.ics.uci.edu/~welling/classnotes/classnotes.html>. The difference between the Kalman filter and smoother is that the latter uses future values as well as past values to estimate the coefficients of interest. We use the Kalman smoother here rather than the filter for two reasons. First, we want to make maximum use of a limited data and the smoother uses more information. Second, given the nature of the problem at hand, the government has to rely on beliefs about future states in order to set policy, so that the future should matter as well as the present in defining policy cyclicity.

set of possible parameters. This is done in such a fashion that the probability of accepting a parameter value is identical to the probability that this parameter value has actually produced the data.

Specifically, in our implementation, we use the classic Metropolis-Hastings (MH) sampler to do MCMC (for an introduction to MCMC and Metropolis-Hastings, see for example Chib and Greenberg (1995)). In MH one starts with arbitrary parameters values. Every iteration one proposes a random change (in our case a small gaussian change) of the parameters. This is what is called the proposal distribution. Subsequently, this change is either accepted or rejected. The probability of acceptance is:

$$p_{accept} = \min \left(1, \frac{p(\text{data}|\text{new_parameters})}{p(\text{data}|\text{previous_parameters})} \right) \quad (1)$$

It is easy to prove that this procedure is actually sampling from the correct posterior distribution over the parameter values.

MCMC algorithms go through two different stages. In the first stage the sampler converges to a probable interpretation of the data in terms of the parameters. This stage is called burn-in and took about 500 iterations in our case. Within these 500 iterations, probabilities increased dramatically and then converged to a stable high level. Afterwards, the MCMC algorithm is exploring the space of relevant parameters. Over 3 runs, we took 10000 samples per run after the end of burn-in. To avoid the autocorrelation that typically characterizes a Markov Chain, we only retain samples every 100 iterations in order to compute the final estimates. From these 3 runs, we thus get a total of 300 essentially uncorrelated samples for each of the five parameters we wish to estimate. Convergence of the Markov chain was assessed comparing the within chain correlation with the across chain correlation. From these 300 samples, we can then directly estimate means and variances of the 5 parameters of interest.

In order to correctly infer the effect of cyclicity on growth in our second stage regressions, we need to determine not only the value of the cyclicity (a_{1it}), but also the uncertainty we have about it. To estimate this uncertainty, or in other words the standard deviation of the cyclicity estimates, it is necessary to consider the relevant sources of uncertainty. Two sources are relevant in our case. One is the uncertainty that is represented by the Kalman smoother that stems from the finite number of noisy observations. The other source of uncertainty is uncertainty about the 5 parameters of the AR(1) processes that are modeled by the MCMC process. To combine them, we use the approximation $\text{variance}_{total} = \text{variance}_{MCMC} + \overline{\text{variance}_{Kalman}}$, where $\overline{\text{variance}_{Kalman}}$ denotes the average variance over the 300 Kalman smoother runs using the 300 samples that we retained from the MCMC estimates of the 5 variances. This approximation becomes correct if the variance as estimated by the Kalman smoother is similar over different runs of the Markov chain, which was a good approximation for our data.

Finally, a full general statistical description of the methods used here can be found in Kording-Marinescu(2006).

Appendix 2: Procyclicality estimates

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1989	Australia		-0.337	-0.693		0.078	0.027		0.050	-0.020
1990	Australia		-0.542	-0.700		0.070	0.026		0.051	-0.024
1991	Australia		-0.780	-0.724		0.061	0.025		0.051	-0.018
1992	Australia		-1.032	-0.786		0.052	0.025		0.051	-0.016
1993	Australia		-1.278	-0.929		0.044	0.024		0.051	-0.023
1994	Australia	-0.224	-1.503	-1.154	0.082	0.037	0.023	0.089	0.051	-0.024
1995	Australia	-1.014	-1.700	-1.222	0.064	0.031	0.023	-0.167	0.052	-0.035
1996	Australia	-0.960	-1.872	-1.292	0.046	0.026	0.023	0.006	0.053	-0.042
1997	Australia	-2.375	-2.025	-1.363	0.040	0.023	0.023	0.109	0.053	-0.049
1998	Australia	-4.056	-2.172	-1.436	-0.035	0.021	0.023	0.163	0.052	-0.055
1999	Australia	-8.856	-2.322	-1.451	-0.047	0.019	0.023	0.104	0.050	-0.072
2000	Australia	-5.228	-2.485	-1.438	0.054	0.017	0.023	-0.279	0.045	-0.041
2001	Australia	-4.397	-2.665	-1.460	0.043	0.017	0.023	-0.220	0.037	-0.039
2002	Australia		-2.864	-1.447		0.016	0.023		0.024	-0.033
2003	Australia		-3.078	-1.452		0.015	0.023		0.007	-0.031
2004	Australia		-3.297	-1.444		0.014	0.023		-0.013	-0.030
2005	Australia		-3.509	-1.443		0.012	0.023		-0.037	-0.029
1972	Austria		-0.849	-0.345		-0.084	0.032		-0.095	0.277
1973	Austria		-0.797	-0.362		-0.075	0.031		-0.117	0.265
1974	Austria		-0.746	-0.342		-0.068	0.031		-0.134	0.237
1975	Austria		-0.698	-0.332		-0.062	0.029		-0.144	0.160
1976	Austria		-0.652	-0.290		-0.056	0.026		-0.147	0.096
1977	Austria	-0.550	-0.608	-0.247	-0.096	-0.049	0.025	-0.129	-0.141	0.029
1978	Austria	-0.655	-0.568	-0.253	-0.022	-0.040	0.024	-0.260	-0.125	0.014
1979	Austria	-0.680	-0.535	-0.259	-0.020	-0.029	0.022	-0.247	-0.100	-0.002
1980	Austria	-0.666	-0.512	-0.250	-0.049	-0.015	0.023	-0.207	-0.066	-0.008
1981	Austria	-0.226	-0.500	-0.233	-0.009	0.000	0.024	-0.207	-0.028	0.010
1982	Austria	-0.065	-0.497	-0.224	-0.101	0.013	0.026	-0.157	0.010	0.031
1983	Austria	-0.200	-0.501	-0.215	-0.154	0.025	0.027	-0.171	0.046	0.073
1984	Austria	-0.650	-0.508	-0.188	-0.062	0.033	0.027	-0.168	0.076	0.106
1985	Austria	-0.304	-0.514	-0.117	0.012	0.040	0.029	-0.036	0.103	0.135
1986	Austria	0.017	-0.515	-0.115	0.044	0.046	0.030	0.155	0.128	0.176
1987	Austria	-0.253	-0.508	-0.079	0.058	0.053	0.034	0.232	0.154	0.233
1988	Austria	-0.397	-0.487	-0.034	0.069	0.062	0.035	0.225	0.182	0.266
1989	Austria	-0.627	-0.449	-0.006	0.105	0.073	0.036	0.307	0.212	0.285
1990	Austria	-0.666	-0.395	0.008	0.103	0.086	0.037	0.309	0.243	0.298
1991	Austria	-0.847	-0.329	-0.048	0.111	0.097	0.036	0.344	0.270	0.311
1992	Austria	-0.604	-0.262	-0.093	0.132	0.104	0.035	0.327	0.287	0.315
1993	Austria	-0.717	-0.199	-0.063	0.058	0.104	0.034	0.293	0.290	0.302
1994	Austria	-1.222	-0.142	-0.034	0.021	0.095	0.033	0.238	0.278	0.290
1995	Austria	0.731	-0.086	-0.005	0.160	0.078	0.032	0.379	0.251	0.276
1996	Austria	0.364	-0.031	0.034	0.108	0.054	0.030	0.170	0.213	0.259
1997	Austria	0.261	0.021	0.064	0.097	0.025	0.029	0.156	0.168	0.241
1998	Austria	0.548	0.063	0.090	-0.004	-0.005	0.027	0.067	0.122	0.222
1999	Austria	0.404	0.091	0.130	-0.074	-0.033	0.025	-0.018	0.081	0.183
2000	Austria	0.013	0.104	0.083	-0.101	-0.054	0.023	-0.022	0.049	0.128
2001	Austria	0.174	0.106	0.090	-0.108	-0.070	0.022	-0.023	0.027	0.107
2002	Austria		0.102	0.087		-0.081	0.021		0.013	0.092
2003	Austria		0.100	0.081		-0.087	0.020		0.004	0.078
2004	Austria		0.102	0.024		-0.091	0.019		-0.001	0.072
2005	Austria		0.112	0.017		-0.094	0.019		-0.004	0.077

Note: the table reports the a_{1it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1972	Germany	-0.201	-0.100		0.090	0.089		0.010	0.219	
1973	Germany	-0.153	-0.165		0.087	0.091		0.012	0.245	
1974	Germany	-0.116	-0.202		0.086	0.095		0.019	0.234	
1975	Germany	-0.091	-0.215		0.086	0.095		0.029	0.181	
1976	Germany	-0.077	-0.197		0.088	0.095		0.042	0.140	
1977	Germany	-0.682	-0.072	-0.167	0.235	0.092	0.097	-0.003	0.057	0.136
1978	Germany	-0.087	-0.078	-0.101	0.116	0.097	0.100	0.071	0.073	0.157
1979	Germany	-0.072	-0.093	-0.134	0.059	0.103	0.101	0.057	0.089	0.165
1980	Germany	-0.204	-0.115	-0.198	0.099	0.110	0.101	0.144	0.104	0.203
1981	Germany	-0.075	-0.144	-0.251	0.196	0.118	0.100	0.348	0.114	0.207
1982	Germany	-0.204	-0.182	-0.314	0.229	0.124	0.100	0.303	0.118	0.203
1983	Germany	-0.316	-0.234	-0.377	0.206	0.128	0.098	0.241	0.113	0.172
1984	Germany	-0.261	-0.307	-0.430	0.178	0.129	0.096	0.177	0.100	0.150
1985	Germany	-0.385	-0.405	-0.496	0.189	0.126	0.095	0.169	0.079	0.128
1986	Germany	-0.312	-0.523	-0.567	0.151	0.120	0.093	0.128	0.052	0.107
1987	Germany	-0.394	-0.653	-0.640	0.173	0.112	0.091	-0.143	0.024	0.082
1988	Germany	-1.510	-0.790	-0.710	0.010	0.102	0.089	-0.521	-0.001	0.056
1989	Germany	-1.074	-0.933	-0.779	0.005	0.092	0.088	-0.172	-0.022	0.028
1990	Germany	-1.480	-1.079	-0.810	-0.039	0.083	0.087	-0.281	-0.034	0.037
1991	Germany	-1.547	-1.217	-0.840	-0.083	0.077	0.086	-0.348	-0.038	0.066
1992	Germany	-2.402	-1.336	-0.818	-0.005	0.074	0.087	-0.298	-0.033	0.140
1993	Germany	-2.516	-1.430	-0.835	0.015	0.074	0.087	-0.281	-0.020	0.166
1994	Germany	-2.237	-1.501	-0.801	0.054	0.076	0.086	-0.201	-0.001	0.168
1995	Germany	-2.451	-1.558	-0.858	0.069	0.080	0.086	-0.104	0.020	0.176
1996	Germany	-1.795	-1.610	-0.783	0.155	0.084	0.085	0.324	0.043	0.184
1997	Germany	-1.265	-1.665	-0.758	0.221	0.087	0.085	0.574	0.067	0.198
1998	Germany	0.084	-1.726	-0.764	0.016	0.090	0.083	0.407	0.092	0.192
1999	Germany	-1.037	-1.793	-0.746	0.122	0.091	0.082	0.258	0.115	0.190
2000	Germany	-2.829	-1.862	-0.772	0.149	0.090	0.082	0.236	0.137	0.191
2001	Germany	-2.534	-1.927	-0.811	0.155	0.088	0.082	0.276	0.155	0.189
2002	Germany		-1.985	-0.851		0.085	0.081		0.169	0.186
2003	Germany		-2.032	-0.860		0.082	0.081		0.180	0.190
2004	Germany		-2.070	-0.881		0.080	0.080		0.186	0.187
2005	Germany		-2.100	-0.899		0.080	0.079		0.191	0.178
1981	Denmark	-2.526	-3.854		0.059	0.042		0.393	-0.533	
1982	Denmark	-2.583	-4.098		0.049	0.042		0.353	-0.498	
1983	Denmark	-2.642	-4.061		0.039	0.042		0.316	-0.343	
1984	Denmark	-2.713	-3.863		0.031	0.040		0.282	-0.263	
1985	Denmark	-2.806	-3.694		0.024	0.039		0.248	-0.207	
1986	Denmark	-2.077	-2.928	-3.476	0.107	0.018	0.036	0.537	0.213	-0.160
1987	Denmark	-1.316	-3.082	-3.587	0.082	0.013	0.037	0.461	0.177	-0.058
1988	Denmark	-2.078	-3.264	-3.664	-0.162	0.010	0.035	0.173	0.139	-0.062
1989	Denmark	-3.931	-3.464	-3.725	-0.013	0.007	0.034	0.067	0.099	-0.059
1990	Denmark	-3.776	-3.661	-3.800	-0.009	0.006	0.033	0.088	0.059	-0.059
1991	Denmark	-3.895	-3.830	-3.884	-0.026	0.006	0.031	0.070	0.024	-0.072
1992	Denmark	-4.584	-3.948	-4.032	0.053	0.006	0.028	0.257	-0.001	-0.074
1993	Denmark	-4.914	-4.004	-4.206	-0.004	0.006	0.029	-0.124	-0.013	-0.083
1994	Denmark	-4.578	-4.001	-3.980	-0.017	0.004	0.028	-0.086	-0.015	-0.065
1995	Denmark	-5.230	-3.946	-3.768	-0.015	0.002	0.028	-0.150	-0.012	-0.051
1996	Denmark	-5.126	-3.846	-3.556	-0.027	0.001	0.028	-0.087	-0.006	-0.036
1997	Denmark	-5.176	-3.704	-3.351	-0.054	0.001	0.028	-0.104	0.000	-0.027
1998	Denmark	-5.557	-3.522	-3.200	0.031	0.003	0.028	-0.084	0.009	-0.005
1999	Denmark	2.109	-3.297	-3.112	0.138	0.006	0.029	0.322	0.020	-0.003
2000	Denmark	2.037	-3.022	-2.856	0.107	0.011	0.030	0.371	0.035	0.010
2001	Denmark	1.633	-2.688	-2.580	0.098	0.016	0.031	0.355	0.053	0.043
2002	Denmark		-2.286	-2.404		0.023	0.031		0.075	0.048
2003	Denmark		-1.816	-2.255		0.030	0.031		0.100	0.055
2004	Denmark		-1.288	-2.259		0.037	0.030		0.127	0.054
2005	Denmark		-0.724	-2.274		0.044	0.030		0.155	0.055

Note: the table reports the a_{1it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1978	Spain		-3.334	-2.535		-0.145	0.084		0.131	-0.850
1979	Spain		-3.121	-2.566		-0.107	0.083		0.123	-0.802
1980	Spain		-2.892	-2.596		-0.070	0.080		0.117	-0.698
1981	Spain		-2.655	-2.577		-0.035	0.078		0.113	-0.546
1982	Spain		-2.419	-2.653		0.001	0.077		0.112	-0.430
1983	Spain	-5.050	-2.194	-2.680	-0.293	0.036	0.086	0.313	0.116	-0.341
1984	Spain	-4.495	-1.993	-2.606	0.297	0.069	0.091	-0.083	0.123	-0.211
1985	Spain	-1.570	-1.829	-2.332	0.295	0.098	0.095	0.078	0.133	-0.118
1986	Spain	-1.193	-1.711	-2.096	0.204	0.119	0.104	0.038	0.143	-0.055
1987	Spain	-1.030	-1.642	-1.976	0.165	0.132	0.112	0.025	0.153	0.001
1988	Spain	-0.813	-1.616	-1.860	0.213	0.139	0.120	0.021	0.162	0.078
1989	Spain	-1.328	-1.625	-1.697	0.135	0.141	0.126	0.172	0.172	0.178
1990	Spain	-1.115	-1.666	-1.668	0.023	0.142	0.129	0.092	0.183	0.236
1991	Spain	-0.883	-1.738	-1.771	0.137	0.142	0.128	0.098	0.197	0.292
1992	Spain	-0.498	-1.847	-1.859	0.162	0.145	0.129	0.250	0.213	0.363
1993	Spain	-1.768	-1.992	-1.941	0.127	0.150	0.130	0.457	0.230	0.427
1994	Spain	-2.388	-2.164	-1.983	0.260	0.157	0.131	0.305	0.246	0.472
1995	Spain	-2.956	-2.348	-2.207	0.262	0.164	0.133	0.300	0.259	0.420
1996	Spain	-2.828	-2.522	-2.350	0.125	0.170	0.135	0.359	0.267	0.421
1997	Spain	-3.200	-2.674	-2.254	0.030	0.174	0.132	0.385	0.272	0.460
1998	Spain	-3.025	-2.796	-2.284	0.134	0.175	0.130	0.289	0.272	0.466
1999	Spain	-2.886	-2.890	-2.259	0.137	0.174	0.129	0.346	0.269	0.464
2000	Spain	-5.457	-2.962	-2.255	0.213	0.171	0.129	-0.004	0.265	0.458
2001	Spain	-6.129	-3.018	-2.261	0.279	0.167	0.129	-0.002	0.258	0.450
2002	Spain		-3.064	-2.257		0.161	0.128		0.250	0.445
2003	Spain		-3.104	-2.253		0.154	0.128		0.242	0.441
2004	Spain		-3.140	-2.249		0.146	0.128		0.232	0.439
2005	Spain		-3.172	-2.244		0.137	0.128		0.222	0.437
1976	Finland		-1.609	-1.340		0.067	0.104		0.560	0.410
1977	Finland		-1.559	-1.354		0.071	0.104		0.493	0.421
1978	Finland		-1.509	-1.374		0.073	0.105		0.424	0.424
1979	Finland		-1.464	-1.384		0.074	0.106		0.357	0.401
1980	Finland		-1.427	-1.394		0.075	0.107		0.298	0.377
1981	Finland	-1.338	-1.402	-1.408	0.125	0.077	0.107	0.695	0.248	0.353
1982	Finland	-0.706	-1.396	-1.417	0.257	0.080	0.107	0.641	0.211	0.323
1983	Finland	-0.289	-1.415	-1.424	0.308	0.086	0.107	0.444	0.186	0.292
1984	Finland	0.095	-1.462	-1.422	0.274	0.095	0.108	-0.004	0.175	0.262
1985	Finland	0.144	-1.538	-1.448	0.235	0.107	0.108	0.005	0.178	0.234
1986	Finland	0.418	-1.635	-1.474	0.263	0.121	0.107	0.071	0.191	0.207
1987	Finland	-0.731	-1.739	-1.505	0.312	0.133	0.107	0.021	0.210	0.174
1988	Finland	-1.668	-1.830	-1.547	0.191	0.142	0.106	0.052	0.229	0.150
1989	Finland	-1.746	-1.892	-1.642	0.212	0.145	0.104	0.070	0.241	0.170
1990	Finland	-3.217	-1.920	-1.819	0.342	0.143	0.113	0.507	0.244	0.241
1991	Finland	-3.296	-1.918	-1.967	0.332	0.137	0.116	0.424	0.239	0.258
1992	Finland	-2.594	-1.893	-2.284	0.206	0.129	0.117	0.372	0.228	0.370
1993	Finland	-1.934	-1.854	-1.738	0.168	0.119	0.114	0.269	0.212	0.342
1994	Finland	-1.779	-1.804	-1.411	0.135	0.109	0.097	0.275	0.193	0.186
1995	Finland	-1.973	-1.748	-1.610	0.105	0.099	0.093	0.218	0.174	0.073
1996	Finland	-1.846	-1.684	-1.389	0.092	0.089	0.090	0.184	0.155	0.043
1997	Finland	-1.429	-1.613	-1.367	0.070	0.079	0.088	0.075	0.137	0.060
1998	Finland	-1.792	-1.534	-1.221	0.066	0.070	0.089	0.050	0.121	0.042
1999	Finland	-1.571	-1.447	-1.026	0.014	0.062	0.089	-0.009	0.106	0.038
2000	Finland	-1.961	-1.355	-0.942	0.017	0.054	0.088	-0.104	0.093	0.024
2001	Finland	-0.511	-1.261	-0.907	-0.073	0.047	0.088	-0.241	0.082	0.025
2002	Finland		-1.168	-0.887		0.040	0.089		0.072	0.029
2003	Finland		-1.077	-0.876		0.034	0.090		0.063	0.047
2004	Finland		-0.988	-0.859		0.029	0.090		0.054	0.067
2005	Finland		-0.897	-0.843		0.025	0.090		0.045	0.069

Note: the table reports the a_{it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1978	France		0.708	0.491		0.015	0.041		0.011	0.145
1979	France		0.615	0.477		0.020	0.042		0.064	0.137
1980	France		0.536	0.453		0.023	0.042		0.108	0.146
1981	France		0.466	0.424		0.023	0.043		0.141	0.150
1982	France		0.400	0.317		0.023	0.044		0.162	0.169
1983	France	2.491	0.328	0.291	-0.041	0.022	0.046	-0.767	0.171	0.207
1984	France	0.437	0.244	0.220	0.174	0.021	0.046	0.085	0.165	0.224
1985	France	0.170	0.147	0.123	0.082	0.019	0.045	0.137	0.143	0.258
1986	France	-0.184	0.043	-0.047	0.027	0.017	0.047	0.121	0.110	0.229
1987	France	-0.027	-0.060	-0.237	0.031	0.014	0.047	0.122	0.071	0.205
1988	France	-0.499	-0.153	-0.357	-0.036	0.012	0.048	0.113	0.032	0.182
1989	France	-0.587	-0.236	-0.484	0.020	0.012	0.049	-0.022	0.000	0.161
1990	France	-0.390	-0.316	-0.631	0.002	0.013	0.051	0.026	-0.021	0.142
1991	France	-0.572	-0.402	-0.725	0.024	0.018	0.053	-0.017	-0.029	0.136
1992	France	-0.900	-0.504	-0.794	0.022	0.027	0.055	-0.026	-0.024	0.127
1993	France	-0.395	-0.624	-0.870	-0.118	0.040	0.058	-0.221	-0.008	0.119
1994	France	1.019	-0.758	-0.942	0.085	0.056	0.058	-0.426	0.016	0.160
1995	France	-2.804	-0.893	-1.067	0.173	0.073	0.059	0.013	0.045	0.161
1996	France	-1.643	-1.016	-1.017	0.206	0.088	0.059	0.114	0.074	0.122
1997	France	-1.452	-1.121	-1.032	0.139	0.100	0.060	0.112	0.100	0.111
1998	France	-0.852	-1.207	-1.116	0.113	0.108	0.058	0.052	0.120	0.146
1999	France	-1.134	-1.281	-1.054	0.096	0.110	0.057	0.134	0.135	0.158
2000	France	-1.431	-1.349	-1.107	0.099	0.109	0.056	0.138	0.146	0.169
2001	France	-1.226	-1.417	-1.170	0.101	0.105	0.055	0.091	0.154	0.178
2002	France		-1.488	-1.221		0.099	0.054		0.160	0.191
2003	France		-1.559	-1.271		0.093	0.053		0.165	0.206
2004	France		-1.625	-1.338		0.086	0.053		0.171	0.199
2005	France		-1.679	-1.374		0.080	0.053		0.177	0.200
1971	United Kingdom		-0.321	-0.749		0.252	0.181		-0.246	-0.595
1972	United Kingdom		-0.174	-0.730		0.238	0.181		-0.170	-0.591
1973	United Kingdom		-0.023	-0.720		0.223	0.181		-0.090	-0.588
1974	United Kingdom		0.123	-0.737		0.208	0.177		-0.010	-0.609
1975	United Kingdom		0.254	-0.752		0.192	0.173		0.064	-0.629
1976	United Kingdom	-0.951	0.357	-0.687	0.255	0.177	0.171	-0.584	0.126	-0.481
1977	United Kingdom	0.451	0.421	-0.610	0.271	0.162	0.168	0.017	0.172	-0.386
1978	United Kingdom	1.665	0.438	-0.509	0.222	0.148	0.162	0.792	0.197	-0.351
1979	United Kingdom	1.749	0.402	-0.407	0.214	0.135	0.157	0.887	0.195	-0.317
1980	United Kingdom	1.763	0.313	-0.309	0.204	0.122	0.151	0.891	0.167	-0.286
1981	United Kingdom	1.926	0.182	-0.076	0.133	0.110	0.148	0.437	0.116	-0.081
1982	United Kingdom	2.063	0.028	0.003	0.176	0.098	0.139	-0.063	0.050	0.037
1983	United Kingdom	1.822	-0.125	-0.226	0.109	0.087	0.131	-0.107	-0.022	0.072
1984	United Kingdom	1.561	-0.262	-0.617	0.062	0.077	0.131	-0.163	-0.089	0.116
1985	United Kingdom	0.652	-0.373	-0.617	0.126	0.067	0.132	-0.363	-0.143	0.122
1986	United Kingdom	0.275	-0.463	-0.781	0.090	0.059	0.133	-0.140	-0.179	0.104
1987	United Kingdom	0.266	-0.548	-0.987	0.037	0.054	0.133	-0.066	-0.194	0.091
1988	United Kingdom	0.393	-0.653	-1.319	-0.083	0.053	0.135	-0.017	-0.186	0.084
1989	United Kingdom	-0.657	-0.805	-1.627	0.068	0.057	0.143	0.258	-0.152	0.089
1990	United Kingdom	-2.160	-1.020	-1.882	0.127	0.066	0.146	0.175	-0.095	0.117
1991	United Kingdom	-2.509	-1.291	-2.080	0.153	0.081	0.148	0.254	-0.023	0.132
1992	United Kingdom	-2.147	-1.591	-2.343	0.279	0.098	0.149	0.330	0.055	0.175
1993	United Kingdom	-2.809	-1.887	-2.608	0.267	0.116	0.151	0.396	0.128	0.238
1994	United Kingdom	-3.360	-2.153	-2.453	0.270	0.133	0.154	0.342	0.195	0.219
1995	United Kingdom	-3.402	-2.377	-2.541	0.234	0.148	0.157	0.352	0.254	0.206
1996	United Kingdom	-4.601	-2.553	-2.507	0.077	0.163	0.160	0.741	0.310	0.190
1997	United Kingdom	-4.421	-2.681	-2.531	0.239	0.178	0.161	0.505	0.362	0.184
1998	United Kingdom	-5.003	-2.767	-2.536	0.184	0.192	0.161	0.738	0.409	0.163
1999	United Kingdom	-5.514	-2.826	-2.536	0.267	0.207	0.161	0.518	0.448	0.141
2000	United Kingdom	-7.900	-2.876	-2.538	0.455	0.222	0.162	0.444	0.474	0.120
2001	United Kingdom	-5.269	-2.932	-2.598	0.574	0.236	0.162	0.159	0.484	0.102
2002	United Kingdom		-3.000	-2.613		0.250	0.162		0.474	0.088
2003	United Kingdom		-3.073	-2.627		0.266	0.161		0.445	0.079
2004	United Kingdom		-3.136	-2.625		0.281	0.162		0.395	0.078
2005	United Kingdom		-3.167	-2.622		0.298	0.162		0.324	0.077

Note: the table reports the a_{1it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption			
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	
1975	Greece			-0.109							
1976	Greece			-0.186							
1977	Greece			-0.269							
1978	Greece			-0.353							
1979	Greece			-0.434							
1980	Greece	-0.167		-0.505							
1981	Greece	-0.404		-0.563							
1982	Greece	-0.851		-0.601							
1983	Greece	-1.101		-0.615							
1984	Greece	-0.719		-0.598							
1985	Greece	-0.578		-0.547							
1986	Greece	0.058		-0.464							
1987	Greece	0.359		-0.365							
1988	Greece	1.565		-0.277							
1989	Greece	-1.524		-0.232							
1990	Greece	-0.215		-0.256							
1991	Greece	-0.211		-0.354							
1992	Greece	-0.920		-0.507							
1993	Greece	-4.638		-0.675							
1994	Greece	-7.648		-0.815							
1995	Greece	-9.259		-0.891							
1996	Greece	-4.887		-0.892							
1997	Greece	-3.114		-0.828							
1998	Greece	1.453		-0.721							
1999	Greece	0.565		-0.591							
2000	Greece	0.426		-0.458							
2001	Greece	0.050		-0.333							
2002	Greece			-0.225							
2003	Greece			-0.140							
2004	Greece			-0.077							
2005	Greece			-0.035							
1978	Ireland			-3.113	-0.142	0.298	0.284		0.434	0.715	
1979	Ireland			-3.136	-0.128	0.280	0.286		0.406	0.756	
1980	Ireland			-3.135	-0.358	0.263	0.285		0.383	0.771	
1981	Ireland			-3.104	-0.462	0.246	0.279		0.365	0.678	
1982	Ireland			-3.038	-0.763	0.230	0.273		0.353	0.598	
1983	Ireland	-3.386		-2.927	-1.073	0.288	0.214	0.267	0.309	0.346	0.514
1984	Ireland	-3.406		-2.764	-1.059	0.215	0.197	0.262	0.235	0.346	0.498
1985	Ireland	-3.531		-2.545	-1.115	0.221	0.181	0.255	0.253	0.353	0.465
1986	Ireland	-3.047		-2.277	-1.279	0.185	0.166	0.250	0.321	0.365	0.410
1987	Ireland	-3.345		-1.982	-1.043	0.193	0.151	0.250	0.339	0.381	0.512
1988	Ireland	-3.404		-1.702	-0.971	0.213	0.138	0.245	0.365	0.396	0.554
1989	Ireland	-2.780		-1.475	-0.959	0.115	0.124	0.238	0.337	0.403	0.539
1990	Ireland	-1.868		-1.311	-0.947	0.081	0.110	0.230	0.326	0.400	0.523
1991	Ireland	-1.953		-1.183	-0.917	0.082	0.096	0.223	0.357	0.390	0.473
1992	Ireland	-0.995		-1.059	-0.894	0.097	0.084	0.217	0.470	0.377	0.424
1993	Ireland	-0.794		-0.925	-0.967	0.048	0.078	0.210	0.419	0.366	0.379
1994	Ireland	-0.795		-0.784	-0.633	0.037	0.078	0.208	0.399	0.357	0.368
1995	Ireland	-0.108		-0.649	-0.575	0.052	0.087	0.209	0.341	0.350	0.388
1996	Ireland	-0.859		-0.531	-0.482	-0.006	0.105	0.210	0.282	0.346	0.396
1997	Ireland	-2.475		-0.439	-0.449	-0.068	0.132	0.212	0.226	0.342	0.389
1998	Ireland	-1.527		-0.378	-0.439	0.177	0.164	0.214	0.122	0.339	0.379
1999	Ireland	0.162		-0.348	-0.418	0.479	0.197	0.216	0.406	0.335	0.366
2000	Ireland	0.010		-0.343	-0.510	0.383	0.227	0.216	0.423	0.332	0.372
2001	Ireland	-0.326		-0.354	-0.407	0.309	0.251	0.215	0.339	0.328	0.405
2002	Ireland			-0.373	-0.408		0.267	0.208		0.325	0.380
2003	Ireland			-0.396	-0.414		0.277	0.206		0.322	0.371
2004	Ireland			-0.419	-0.416		0.281	0.206		0.319	0.373
2005	Ireland			-0.440	-0.411		0.281	0.206		0.318	0.377

Note: the table reports the a_{it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1981	Iceland		-1.908	-1.588		0.023	0.110		-0.310	-0.392
1982	Iceland		-1.864	-1.442		0.029	0.109		-0.258	-0.370
1983	Iceland		-1.815	-1.479		0.036	0.107		-0.201	-0.367
1984	Iceland		-1.759	-1.445		0.045	0.109		-0.136	-0.201
1985	Iceland		-1.697	-1.364		0.054	0.109		-0.061	-0.163
1986	Iceland	-1.773	-1.630	-1.309	0.055	0.065	0.112	-0.164	0.020	-0.026
1987	Iceland	-1.394	-1.565	-1.234	0.043	0.075	0.116	-0.160	0.104	0.117
1988	Iceland	-1.710	-1.507	-1.207	0.064	0.084	0.119	-0.176	0.183	0.160
1989	Iceland	-1.545	-1.464	-1.181	0.167	0.092	0.121	0.201	0.252	0.205
1990	Iceland	-1.109	-1.437	-1.161	0.092	0.098	0.123	0.279	0.304	0.251
1991	Iceland	-0.786	-1.428	-1.146	0.058	0.101	0.125	0.628	0.336	0.294
1992	Iceland	-1.123	-1.433	-1.248	0.133	0.102	0.127	0.610	0.351	0.333
1993	Iceland	-1.384	-1.444	-1.406	0.146	0.101	0.124	0.415	0.351	0.231
1994	Iceland	-1.587	-1.454	-1.425	0.199	0.097	0.133	0.367	0.339	0.197
1995	Iceland	-1.731	-1.457	-1.336	0.169	0.093	0.142	0.344	0.320	0.144
1996	Iceland	-1.615	-1.448	-1.169	0.079	0.088	0.139	0.286	0.297	0.122
1997	Iceland	-1.272	-1.424	-1.001	0.100	0.085	0.140	0.297	0.276	0.103
1998	Iceland	-1.667	-1.385	-0.828	0.076	0.085	0.140	0.240	0.260	0.078
1999	Iceland	-2.662	-1.332	-0.655	0.323	0.089	0.137	0.260	0.253	0.020
2000	Iceland	-0.927	-1.260	-0.377	0.510	0.101	0.134	0.413	0.256	-0.040
2001	Iceland	0.596	-1.162	-0.005	0.408	0.121	0.137	0.552	0.269	-0.074
2002	Iceland		-1.026	-0.045		0.149	0.135		0.291	-0.094
2003	Iceland		-0.844	-0.081		0.184	0.134		0.319	-0.118
2004	Iceland		-0.612	-0.092		0.224	0.133		0.351	-0.137
2005	Iceland		-0.334	-0.086		0.267	0.132		0.386	-0.138
1965	Italy		1.193	0.280		0.148	0.103		0.066	-0.164
1966	Italy		1.162	0.289		0.139	0.102		0.057	-0.125
1967	Italy		1.141	0.233		0.129	0.101		0.054	-0.123
1968	Italy		1.125	0.176		0.119	0.100		0.059	-0.121
1969	Italy		1.107	0.104		0.110	0.099		0.074	-0.117
1970	Italy	0.282	1.078	0.062	0.170	0.102	0.099	-0.028	0.100	-0.098
1971	Italy	1.538	1.026	0.165	0.027	0.095	0.097	0.311	0.137	-0.020
1972	Italy	1.753	0.940	0.251	0.066	0.090	0.096	0.287	0.183	0.010
1973	Italy	1.363	0.814	0.337	0.015	0.087	0.095	-0.308	0.233	0.041
1974	Italy	0.928	0.647	0.368	-0.050	0.086	0.095	-0.262	0.279	0.084
1975	Italy	-0.661	0.451	0.282	-0.036	0.084	0.094	-0.026	0.313	0.113
1976	Italy	-0.335	0.243	0.215	-0.026	0.082	0.095	0.124	0.330	0.131
1977	Italy	-0.373	0.049	0.152	0.031	0.080	0.096	0.212	0.329	0.153
1978	Italy	-0.323	-0.111	0.092	0.046	0.076	0.098	0.265	0.316	0.181
1979	Italy	-0.247	-0.224	0.027	0.091	0.074	0.099	0.342	0.298	0.210
1980	Italy	-0.404	-0.291	0.018	0.087	0.072	0.102	0.458	0.284	0.239
1981	Italy	-0.345	-0.317	0.037	0.154	0.072	0.103	0.373	0.275	0.291
1982	Italy	-0.450	-0.309	0.084	0.120	0.073	0.102	0.248	0.273	0.293
1983	Italy	-0.278	-0.268	0.140	0.019	0.075	0.101	0.296	0.277	0.283
1984	Italy	-0.470	-0.191	0.162	0.054	0.078	0.100	0.367	0.284	0.308
1985	Italy	-0.735	-0.073	0.183	0.104	0.081	0.099	0.235	0.293	0.304
1986	Italy	-0.269	0.084	0.295	0.144	0.083	0.099	0.456	0.302	0.317
1987	Italy	0.283	0.273	0.365	0.025	0.086	0.099	0.262	0.311	0.279
1988	Italy	0.271	0.476	0.462	0.078	0.088	0.100	0.458	0.317	0.269
1989	Italy	1.928	0.666	0.566	0.091	0.091	0.100	0.316	0.321	0.253
1990	Italy	1.249	0.818	0.753	0.073	0.095	0.101	0.344	0.324	0.291
1991	Italy	1.771	0.910	0.814	0.130	0.101	0.101	0.314	0.325	0.286
1992	Italy	2.100	0.929	0.897	0.184	0.107	0.102	0.123	0.327	0.278
1993	Italy	1.297	0.866	1.017	0.290	0.113	0.102	0.392	0.330	0.257
1994	Italy	0.293	0.722	0.758	0.362	0.118	0.100	0.569	0.334	0.261
1995	Italy	-1.579	0.501	0.725	0.345	0.121	0.097	0.458	0.341	0.248
1996	Italy	-1.670	0.220	0.676	0.356	0.120	0.094	0.526	0.350	0.208
1997	Italy	-2.088	-0.093	0.679	0.317	0.115	0.091	0.518	0.361	0.214
1998	Italy	-2.217	-0.399	0.624	0.352	0.105	0.089	0.553	0.374	0.229
1999	Italy	-3.625	-0.661	0.620	-0.011	0.092	0.088	0.417	0.385	0.235
2000	Italy	-1.371	-0.861	0.555	-0.035	0.075	0.087	0.259	0.392	0.240
2001	Italy	-1.430	-1.001	0.489	-0.026	0.058	0.085	0.340	0.394	0.243
2002	Italy		-1.100	0.430		0.041	0.084		0.391	0.242
2003	Italy		-1.171	0.374		0.026	0.081		0.384	0.231
2004	Italy		-1.224	0.322		0.011	0.083		0.376	0.245
2005	Italy		-1.262	0.296		-0.002	0.083		0.367	0.271

Note: the table reports the a_{lit} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1971	Japan		-1.564	-1.220		0.399	0.061		-0.368	-0.227
1972	Japan		-1.552	-1.242		0.399	0.061		-0.353	-0.218
1973	Japan		-1.539	-1.280		0.399	0.062		-0.334	-0.207
1974	Japan		-1.524	-1.287		0.399	0.060		-0.313	-0.217
1975	Japan		-1.502	-1.336		0.397	0.057		-0.289	-0.187
1976	Japan	-1.214	-1.475	-1.427	0.434	0.394	0.054	-0.411	-0.261	-0.135
1977	Japan	-0.512	-1.451	-1.449	0.421	0.388	0.049	-0.357	-0.231	-0.125
1978	Japan	-0.459	-1.448	-1.466	0.418	0.377	0.046	-0.291	-0.196	-0.112
1979	Japan	2.733	-1.487	-1.452	0.448	0.360	0.043	-0.512	-0.158	-0.101
1980	Japan	3.466	-1.589	-1.435	0.511	0.336	0.040	-0.672	-0.117	-0.085
1981	Japan	0.233	-1.760	-1.436	0.503	0.305	0.038	-0.058	-0.078	-0.076
1982	Japan	-0.367	-1.982	-1.439	0.271	0.272	0.036	-0.080	-0.043	-0.066
1983	Japan	-2.771	-2.232	-1.444	-0.063	0.237	0.034	0.083	-0.016	-0.055
1984	Japan	-2.634	-2.490	-1.444	0.366	0.202	0.031	-0.032	0.002	-0.044
1985	Japan	-2.599	-2.747	-1.458	0.194	0.170	0.027	0.063	0.012	-0.035
1986	Japan	-2.188	-2.984	-1.486	0.115	0.141	0.023	0.052	0.015	-0.025
1987	Japan	-2.801	-3.158	-1.464	0.106	0.116	0.019	0.061	0.016	-0.013
1988	Japan	-2.680	-3.212	-1.506	0.174	0.092	0.016	0.045	0.018	-0.004
1989	Japan	-3.372	-3.127	-1.544	0.014	0.064	0.014	0.113	0.024	0.007
1990	Japan	-3.856	-2.939	-1.572	-0.172	0.029	0.011	0.138	0.032	0.010
1991	Japan	-3.421	-2.713	-1.603	-0.289	-0.011	0.008	0.148	0.039	0.019
1992	Japan	-2.212	-2.496	-1.490	-0.133	-0.050	0.006	0.126	0.043	0.017
1993	Japan	-2.445	-2.304	-1.412	-0.158	-0.086	0.003	0.103	0.042	0.013
1994	Japan	-2.636	-2.131	-1.332	-0.200	-0.115	0.001	0.090	0.036	0.008
1995	Japan	-2.241	-1.958	-1.270	-0.268	-0.137	-0.002	0.054	0.028	0.003
1996	Japan	-2.673	-1.763	-1.201	-0.366	-0.150	-0.006	0.066	0.019	0.000
1997	Japan	-0.584	-1.536	-1.140	-0.471	-0.155	-0.010	-0.048	0.009	0.001
1998	Japan	1.005	-1.277	-1.090	-0.588	-0.152	-0.012	-0.075	0.002	0.003
1999	Japan	0.677	-0.998	-1.015	-0.476	-0.142	-0.013	-0.183	-0.003	0.005
2000	Japan	0.015	-0.714	-0.872	-0.355	-0.127	-0.012	-0.141	-0.004	0.008
2001	Japan	0.039	-0.440	-0.731	-0.135	-0.110	-0.012	-0.016	-0.001	0.012
2002	Japan		-0.186	-0.618		-0.091	-0.011		0.004	0.025
2003	Japan		0.044	-0.637		-0.072	-0.009		0.011	0.053
2004	Japan		0.247	-0.634		-0.054	-0.008		0.021	0.059
2005	Japan		0.425	-0.631		-0.035	-0.007		0.031	0.066
1972	Netherlands		-0.486	-0.469		0.113	0.080		0.368	0.176
1973	Netherlands		-0.528	-0.471		0.115	0.079		0.371	0.174
1974	Netherlands		-0.570	-0.480		0.115	0.079		0.375	0.182
1975	Netherlands		-0.610	-0.490		0.114	0.079		0.376	0.153
1976	Netherlands		-0.647	-0.497		0.111	0.080		0.377	0.156
1977	Netherlands	-0.244	-0.678	-0.506	0.075	0.108	0.081	0.255	0.376	0.163
1978	Netherlands	-0.199	-0.705	-0.522	0.072	0.103	0.082	-0.132	0.373	0.169
1979	Netherlands	-0.137	-0.725	-0.537	0.085	0.098	0.083	-0.048	0.370	0.175
1980	Netherlands	-0.446	-0.737	-0.553	0.169	0.093	0.084	-0.123	0.365	0.182
1981	Netherlands	-0.271	-0.739	-0.563	0.133	0.088	0.085	0.219	0.360	0.181
1982	Netherlands	-0.744	-0.728	-0.590	0.094	0.083	0.086	0.132	0.353	0.162
1983	Netherlands	-1.008	-0.702	-0.638	0.086	0.079	0.085	0.284	0.346	0.228
1984	Netherlands	-0.842	-0.662	-0.616	0.083	0.076	0.083	0.239	0.336	0.293
1985	Netherlands	-0.677	-0.611	-0.601	0.100	0.074	0.084	0.172	0.325	0.241
1986	Netherlands	-0.088	-0.552	-0.522	0.045	0.072	0.085	0.182	0.313	0.237
1987	Netherlands	0.705	-0.494	-0.454	0.011	0.071	0.085	0.208	0.301	0.231
1988	Netherlands	0.718	-0.450	-0.440	0.015	0.071	0.086	0.144	0.289	0.245
1989	Netherlands	0.458	-0.439	-0.408	0.077	0.072	0.087	0.234	0.281	0.249
1990	Netherlands	0.709	-0.476	-0.378	0.075	0.073	0.087	0.221	0.276	0.251
1991	Netherlands	0.749	-0.559	-0.469	0.071	0.076	0.088	0.212	0.274	0.280
1992	Netherlands	0.658	-0.666	-0.560	0.046	0.078	0.090	0.205	0.276	0.295
1993	Netherlands	-0.055	-0.771	-0.654	0.056	0.081	0.092	0.256	0.281	0.303
1994	Netherlands	1.784	-0.879	-0.753	0.080	0.084	0.094	0.135	0.289	0.308
1995	Netherlands	1.682	-1.010	-0.862	0.080	0.088	0.095	0.186	0.302	0.311
1996	Netherlands	0.767	-1.170	-0.984	0.097	0.094	0.096	0.329	0.319	0.311
1997	Netherlands	-1.593	-1.341	-1.102	0.058	0.101	0.097	0.017	0.338	0.323
1998	Netherlands	-2.005	-1.498	-1.174	-0.022	0.109	0.098	0.077	0.358	0.314
1999	Netherlands	-1.746	-1.627	-1.283	0.093	0.116	0.099	0.280	0.376	0.303
2000	Netherlands	-1.975	-1.725	-1.408	0.122	0.122	0.101	0.364	0.392	0.323
2001	Netherlands	-1.891	-1.796	-1.416	0.125	0.128	0.103	0.428	0.406	0.366
2002	Netherlands		-1.847	-1.454		0.133	0.104		0.418	0.385
2003	Netherlands		-1.881	-1.492		0.137	0.106		0.427	0.403
2004	Netherlands		-1.904	-1.535		0.141	0.106		0.435	0.414
2005	Netherlands		-1.918	-1.554		0.145	0.105		0.442	0.428

Note: the table reports the a_{lit} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years

rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.
Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1971	Norway		1.080	1.346		0.048	0.092		0.174	0.206
1972	Norway		1.244	1.266		0.067	0.091		0.194	0.144
1973	Norway		1.396	1.275		0.083	0.086		0.211	0.147
1974	Norway		1.531	1.193		0.096	0.086		0.225	0.176
1975	Norway		1.645	1.207		0.106	0.086		0.237	0.248
1976	Norway	0.564	1.738	1.056	-0.019	0.113	0.081	0.089	0.245	0.220
1977	Norway	1.943	1.812	1.447	0.124	0.117	0.080	0.121	0.251	0.154
1978	Norway	1.958	1.869	1.391	0.139	0.120	0.077	0.168	0.254	0.142
1979	Norway	1.943	1.913	1.094	0.124	0.120	0.073	0.219	0.255	0.103
1980	Norway	1.650	1.941	1.062	0.114	0.119	0.073	0.261	0.252	0.154
1981	Norway	1.465	1.951	1.083	0.074	0.118	0.073	0.150	0.247	0.167
1982	Norway	2.584	1.937	1.129	0.098	0.115	0.074	0.037	0.238	0.158
1983	Norway	1.610	1.893	1.168	0.022	0.113	0.075	-0.111	0.227	0.152
1984	Norway	2.697	1.813	1.154	0.084	0.110	0.076	0.202	0.215	0.158
1985	Norway	1.498	1.700	1.153	0.057	0.108	0.078	0.177	0.202	0.154
1986	Norway	1.275	1.555	1.125	0.090	0.106	0.080	0.183	0.190	0.161
1987	Norway	1.488	1.384	0.830	0.103	0.104	0.080	0.099	0.178	0.205
1988	Norway	1.475	1.189	0.698	0.109	0.102	0.080	0.094	0.164	0.193
1989	Norway	1.268	0.975	0.539	0.124	0.100	0.080	0.112	0.147	0.140
1990	Norway	1.243	0.749	0.589	0.132	0.096	0.082	0.149	0.126	0.076
1991	Norway	2.147	0.524	0.450	0.163	0.092	0.081	0.193	0.102	0.058
1992	Norway	-0.126	0.313	-0.061	0.106	0.088	0.084	0.132	0.078	0.087
1993	Norway	-0.008	0.121	-0.635	0.119	0.083	0.086	-0.109	0.056	0.141
1994	Norway	-0.585	-0.053	-0.830	0.099	0.080	0.084	-0.046	0.040	0.176
1995	Norway	-0.437	-0.218	-1.019	0.084	0.079	0.081	-0.090	0.029	0.191
1996	Norway	-1.136	-0.384	-1.201	0.052	0.079	0.079	-0.035	0.024	0.202
1997	Norway	-2.887	-0.558	-1.374	0.165	0.081	0.077	0.078	0.024	0.211
1998	Norway	-6.092	-0.749	-1.478	0.372	0.085	0.073	0.225	0.030	0.232
1999	Norway	-7.026	-0.966	-1.458	0.326	0.090	0.069	0.190	0.041	0.227
2000	Norway	-7.278	-1.215	-1.384	0.384	0.096	0.067	0.249	0.058	0.244
2001	Norway	-5.882	-1.501	-1.375	0.296	0.102	0.066	0.326	0.082	0.284
2002	Norway		-1.817	-1.327		0.107	0.065		0.115	0.300
2003	Norway		-2.146	-1.270		0.109	0.064		0.156	0.315
2004	Norway		-2.456	-1.254		0.107	0.064		0.206	0.316
2005	Norway		-2.703	-1.238		0.098	0.064		0.261	0.318
1995	New Zealand		-0.303	-0.252		-0.091	-0.105		0.584	0.536
1996	New Zealand		-0.396	-0.252		-0.094	-0.106		0.550	0.530
1997	New Zealand		-0.494	-0.208		-0.096	-0.106		0.516	0.543
1998	New Zealand		-0.592	-0.172		-0.099	-0.106		0.484	0.552
1999	New Zealand		-0.687	-0.150		-0.101	-0.107		0.454	0.503
2000	New Zealand	-0.998	-0.771	-0.136	-0.077	-0.104	-0.108	0.488	0.427	0.492
2001	New Zealand	-0.728	-0.838	-0.127	-0.094	-0.108	-0.108	0.511	0.403	0.505
2002	New Zealand		-0.884	-0.135		-0.113	-0.108		0.384	0.509
2003	New Zealand		-0.905	-0.143		-0.118	-0.108		0.371	0.509
2004	New Zealand		-0.898	-0.145		-0.124	-0.107		0.364	0.520
2005	New Zealand		-0.866	-0.145		-0.132	-0.107		0.365	0.520

Note: the table reports the a_{it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1971	Portugal		-2.163	-1.696						
1972	Portugal		-2.226	-1.632						
1973	Portugal		-2.278	-1.508						
1974	Portugal		-2.310	-1.616						
1975	Portugal		-2.308	-1.670						
1976	Portugal	-2.044	-2.259	-1.689						
1977	Portugal	-2.048	-2.162	-1.710						
1978	Portugal	-2.060	-2.027	-1.759	0.053	0.049		-0.003	0.275	
1979	Portugal	-4.116	-1.873	-1.809	0.053	0.048		0.018	0.276	
1980	Portugal	-5.985	-1.718	-1.618	0.052	0.049		0.032	0.301	
1981	Portugal	-6.769	-1.568	-1.560	0.051	0.051		0.042	0.260	
1982	Portugal	-7.653	-1.421	-1.502		0.049	0.054		0.049	0.206
1983	Portugal	-2.455	-1.274	-1.489	0.091	0.048	0.054	-0.317	0.054	0.137
1984	Portugal	-1.090	-1.122	-1.364	0.104	0.048	0.050	-0.029	0.059	0.100
1985	Portugal	-0.108	-0.971	-1.103	0.026	0.046	0.043	0.116	0.066	0.032
1986	Portugal	0.046	-0.826	-0.752	0.027	0.045	0.038	-0.049	0.076	0.083
1987	Portugal	-0.262	-0.693	-1.002	-0.024	0.043	0.035	-0.059	0.091	0.235
1988	Portugal	0.383	-0.576	-0.948	-0.043	0.040	0.033	0.109	0.113	0.276
1989	Portugal	0.581	-0.479	-0.874	-0.030	0.038	0.032	0.434	0.141	0.323
1990	Portugal	0.248	-0.404	-0.830	0.075	0.034	0.032	0.407	0.173	0.368
1991	Portugal	0.194	-0.358	-0.765	0.055	0.031	0.033	0.395	0.206	0.415
1992	Portugal	-0.568	-0.346	-1.029	0.057	0.029	0.037	0.583	0.234	0.356
1993	Portugal	-0.583	-0.375	-1.039	0.122	0.027	0.041	0.512	0.253	0.349
1994	Portugal	-0.808	-0.444	-1.046	0.091	0.028	0.043	0.409	0.262	0.333
1995	Portugal	-0.810	-0.548	-1.062	0.073	0.034	0.042	0.318	0.263	0.305
1996	Portugal	-0.827	-0.677	-1.028	0.033	0.047	0.043	0.250	0.264	0.300
1997	Portugal	-1.318	-0.821	-0.994	0.037	0.068	0.044	0.142	0.270	0.296
1998	Portugal	-1.115	-0.969	-0.952	0.089	0.097	0.044	0.152	0.283	0.293
1999	Portugal	-1.470	-1.111	-0.862	0.168	0.133	0.047	0.275	0.305	0.321
2000	Portugal	-1.616	-1.240	-0.805	0.444	0.172	0.048	0.363	0.330	0.338
2001	Portugal	-1.359	-1.350	-0.702	0.489	0.213	0.051	0.510	0.356	0.331
2002	Portugal		-1.440	-0.641		0.253	0.053		0.378	0.331
2003	Portugal		-1.512	-0.562		0.291	0.054		0.396	0.331
2004	Portugal		-1.571	-0.531		0.327	0.059		0.409	0.316
2005	Portugal		-1.624	-0.549		0.362	0.056		0.418	0.326
1971	Sweden		-0.856	-0.936		-0.061	-0.001		0.010	-0.003
1972	Sweden		-1.016	-0.929		-0.068	0.000		-0.014	-0.004
1973	Sweden		-1.181	-0.913		-0.073	0.000		-0.030	0.005
1974	Sweden		-1.348	-0.936		-0.073	0.002		-0.040	0.019
1975	Sweden		-1.513	-1.021		-0.070	0.004		-0.043	0.073
1976	Sweden	-1.092	-1.670	-1.180	-0.117	-0.064	0.003	-0.120	-0.040	0.046
1977	Sweden	-0.861	-1.814	-1.210	-0.133	-0.056	0.004	-0.125	-0.032	0.017
1978	Sweden	-0.961	-1.940	-1.408	-0.137	-0.047	0.007	-0.161	-0.022	0.062
1979	Sweden	-1.437	-2.048	-1.644	-0.074	-0.037	0.011	-0.023	-0.011	0.056
1980	Sweden	-1.477	-2.138	-1.882	-0.058	-0.027	0.016	0.070	0.002	0.070
1981	Sweden	-2.492	-2.216	-2.070	-0.065	-0.018	0.020	0.033	0.016	0.105
1982	Sweden	-2.010	-2.286	-2.253	-0.037	-0.008	0.024	0.029	0.033	0.105
1983	Sweden	-2.379	-2.353	-2.326	-0.032	0.003	0.026	-0.008	0.053	0.061
1984	Sweden	-2.878	-2.416	-2.463	-0.017	0.016	0.030	-0.117	0.076	0.058
1985	Sweden	-2.368	-2.480	-2.593	0.068	0.031	0.034	0.068	0.100	0.055
1986	Sweden	-2.682	-2.546	-2.701	0.039	0.046	0.038	0.090	0.119	0.054
1987	Sweden	-3.294	-2.621	-2.897	0.074	0.062	0.043	0.093	0.132	0.040
1988	Sweden	-3.769	-2.708	-2.965	0.084	0.077	0.049	0.236	0.138	0.058
1989	Sweden	-2.929	-2.805	-3.025	0.085	0.090	0.052	0.332	0.138	0.163
1990	Sweden	-2.734	-2.906	-3.171	0.088	0.101	0.047	0.381	0.136	0.260
1991	Sweden	-2.918	-3.000	-3.062	0.128	0.108	0.046	0.356	0.133	0.270
1992	Sweden	-3.599	-3.079	-2.940	0.217	0.113	0.044	0.104	0.130	0.281
1993	Sweden	-3.378	-3.138	-2.208	0.199	0.113	0.039	0.156	0.124	0.202
1994	Sweden	-3.461	-3.177	-2.430	0.196	0.109	0.035	0.294	0.114	0.156
1995	Sweden	-3.646	-3.199	-2.425	0.098	0.103	0.037	0.327	0.100	0.101
1996	Sweden	-2.931	-3.211	-2.394	0.058	0.094	0.041	-0.004	0.084	0.027
1997	Sweden	-3.194	-3.216	-2.365	0.071	0.083	0.040	0.046	0.066	0.028
1998	Sweden	-2.339	-3.220	-2.477	0.039	0.073	0.037	0.075	0.051	-0.041
1999	Sweden	-3.403	-3.224	-2.536	0.034	0.063	0.036	0.007	0.039	-0.078
2000	Sweden	-3.340	-3.228	-2.552	0.086	0.055	0.034	0.001	0.031	-0.131
2001	Sweden	-3.294	-3.231	-2.550	0.098	0.048	0.035	-0.011	0.026	-0.117
2002	Sweden		-3.235	-2.551		0.042	0.036		0.022	-0.101
2003	Sweden		-3.238	-2.552		0.038	0.036		0.018	-0.091
2004	Sweden		-3.241	-2.553		0.034	0.036		0.015	-0.082
2005	Sweden		-3.241	-2.551		0.031	0.036		0.010	-0.082

Note: the table reports the a_{lit} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years

rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.
Source: OECD Economic Outlook.

Year	Country	Procyclicality of public debt			Procyclicality of public investment			Procyclicality of public consumption		
		10YRW	WRW	AR(1)	10YRW	WRW	AR(1)	10YRW	WRW	AR(1)
1964	United States	-1.093	-0.962		0.170	0.082		0.273	0.183	
1965	United States	-1.144	-0.964		0.156	0.082		0.224	0.184	
1966	United States	-1.187	-0.981		0.141	0.082		0.175	0.189	
1967	United States	-1.219	-1.016		0.126	0.079		0.129	0.151	
1968	United States	-1.233	-1.045		0.112	0.076		0.088	0.097	
1969	United States	-1.052	-1.226	-1.066	0.250	0.098	0.074	0.185	0.053	0.053
1970	United States	-1.043	-1.196	-1.056	0.216	0.084	0.073	0.037	0.026	0.018
1971	United States	-1.431	-1.141	-1.054	0.132	0.072	0.072	-0.076	0.005	-0.020
1972	United States	-1.464	-1.064	-1.029	0.129	0.060	0.070	-0.066	-0.010	-0.056
1973	United States	-1.426	-0.970	-1.002	0.126	0.049	0.069	-0.040	-0.019	-0.092
1974	United States	-1.075	-0.869	-0.981	0.065	0.038	0.067	-0.177	-0.026	-0.102
1975	United States	-0.602	-0.771	-0.972	0.018	0.029	0.066	-0.355	-0.030	-0.104
1976	United States	-0.378	-0.679	-0.900	0.060	0.024	0.066	-0.412	-0.035	-0.086
1977	United States	0.732	-0.596	-0.827	0.121	0.024	0.066	-0.329	-0.037	-0.085
1978	United States	-0.358	-0.529	-0.753	0.055	0.028	0.066	-0.097	-0.035	-0.086
1979	United States	-0.215	-0.479	-0.728	0.035	0.034	0.066	-0.083	-0.030	-0.080
1980	United States	-0.206	-0.448	-0.720	0.047	0.039	0.065	-0.058	-0.023	-0.073
1981	United States	-0.196	-0.432	-0.721	0.058	0.044	0.065	-0.011	-0.014	-0.054
1982	United States	-0.297	-0.428	-0.763	0.054	0.049	0.064	0.010	-0.005	-0.020
1983	United States	-0.295	-0.433	-0.822	0.074	0.052	0.063	0.010	0.004	0.010
1984	United States	-0.453	-0.445	-0.904	0.080	0.056	0.063	0.060	0.012	0.012
1985	United States	-0.677	-0.466	-0.973	0.064	0.059	0.063	0.162	0.020	0.012
1986	United States	-0.496	-0.498	-1.023	0.048	0.062	0.063	0.101	0.028	0.016
1987	United States	-0.480	-0.550	-1.078	0.072	0.066	0.064	-0.037	0.036	0.021
1988	United States	-0.663	-0.631	-1.139	0.068	0.070	0.064	0.083	0.042	0.024
1989	United States	-1.362	-0.752	-1.208	0.039	0.074	0.065	0.062	0.047	0.030
1990	United States	-2.021	-0.920	-1.288	0.043	0.078	0.065	-0.085	0.050	0.033
1991	United States	-1.979	-1.139	-1.366	0.084	0.082	0.066	-0.063	0.051	0.034
1992	United States	-2.103	-1.398	-1.412	0.111	0.084	0.067	-0.049	0.050	0.048
1993	United States	-2.319	-1.674	-1.464	0.102	0.087	0.067	-0.040	0.047	0.052
1994	United States	-2.528	-1.934	-1.516	0.101	0.088	0.067	-0.061	0.044	0.046
1995	United States	-2.882	-2.145	-1.567	0.114	0.089	0.067	-0.018	0.043	0.040
1996	United States	-3.464	-2.286	-1.630	0.092	0.089	0.066	-0.053	0.044	0.033
1997	United States	-2.228	-2.352	-1.694	0.089	0.087	0.066	0.159	0.046	0.025
1998	United States	-3.044	-2.346	-1.759	0.084	0.083	0.066	0.146	0.047	0.017
1999	United States	-2.869	-2.275	-1.822	0.074	0.079	0.066	0.146	0.045	0.016
2000	United States	-1.790	-2.148	-1.911	0.054	0.074	0.066	0.079	0.040	0.005
2001	United States	-1.833	-1.984	-1.923	0.064	0.070	0.066	0.039	0.032	-0.010
2002	United States		-1.807	-1.958		0.065	0.066		0.022	-0.024
2003	United States		-1.637	-1.979		0.062	0.066		0.009	-0.030
2004	United States		-1.487	-1.987		0.059	0.066		-0.006	-0.029
2005	United States		-1.363	-1.986		0.058	0.066		-0.022	-0.030

Note: the table reports the a_{1it} coefficients, i.e. the coefficients on the output gap composite variable (see equation 1), using various estimation techniques. 10YRW stands for the 10-years rolling window OLS method, WRW for the Gaussian-weighted rolling window OLS method and AR(1) for the AR(1) MCMC method.

Source: OECD Economic Outlook.

Appendix 3: Further results for the second stage regressions

Table 1: The effect of public debt procyclicality on growth, 10-years rolling window method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government debt)	-0.002 (0.002)	-0.004 (0.003)	0.000 (0.002)	-0.001 (0.002)	0.002 (0.002)
lag(Private credit/GDP)	-0.012 (0.007)*	0.006 (0.011)	-0.008 (0.010)	-0.016 (0.009)*	-0.032 (0.008)***
lag(Procyclicality of government debt*Private credit/GDP)	-0.001 (0.002)	0.002 (0.003)	-0.001 (0.002)	0.002 (0.002)	-0.002 (0.002)
lag(log (real GDP per capita))	-0.002 (0.001)*	-0.129 (0.023)***	-0.198 (0.021)***	-0.246 (0.026)***	-0.328 (0.030)***
Average years of schooling for the population over 25 years old	0.000 (0.001)	0.009 (0.005)*	0.010 (0.004)***	-0.001 (0.004)	0.002 (0.003)
Openness	0.000 (0.000)**	0.001 (0.000)***	0.002 (0.000)***	0.000 (0.000)	0.000 (0.000)*
Inflation	-0.086 (0.042)**	-0.166 (0.043)***	-0.129 (0.036)***	-0.011 (0.030)	-0.011 (0.036)
Population growth	-0.573 (0.371)	-0.428 (0.797)	-1.641 (0.433)***	-0.412 (0.344)	-1.135 (0.381)***
Government share of GDP (in %)	-0.001 (0.000)***	-0.001 (0.000)*	-0.001 (0.000)**	0.000 (0.000)	0.000 (0.000)
Investment/GDP (in%)	0.002 (0.001)***	0.005 (0.001)***	0.005 (0.001)***	0.004 (0.001)***	0.004 (0.001)***
Constant	-0.013 (0.028)	-1.252 (0.226)***	-1.794 (0.112)***	-2.149 (0.225)***	-2.781 (0.156)***
Observations	385	366	366	366	366
R-squared	0.22	0.46		0.77	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 2: The effect of public debt procyclicality on growth, 10-years rolling window method, without controlling for lag(log(real GDP per capita))

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government debt)	-0.001 (0.002)	-0.008 (0.003)***	-0.001 (0.002)	-0.003 (0.003)	-0.000 (0.002)
lag(Private credit/GDP)	-0.010 (0.007)	-0.002 (0.011)	-0.026 (0.010)**	0.004 (0.011)	-0.010 (0.009)
lag(Procyclicality of government debt*Private credit/GDP)	-0.001 (0.002)	0.007 (0.003)**	0.001 (0.002)	0.005 (0.003)*	0.000 (0.002)
Average years of schooling for the population over 25 years old	-0.000 (0.001)	-0.002 (0.004)	-0.007 (0.003)*	-0.001 (0.004)	0.002 (0.004)
Openness	-0.000 (0.000)**	0.000 (0.000)	0.001 (0.000)***	-0.001 (0.000)*	0.000 (0.000)
Inflation	-0.093 (0.043)**	-0.152 (0.045)***	-0.104 (0.039)***	-0.089 (0.042)**	-0.104 (0.040)***
Population growth	-0.637 (0.386)*	-1.081 (0.838)	-2.084 (0.466)***	-0.440 (0.530)	-1.130 (0.439)**
Government share of GDP (in %)	-0.001 (0.000)**	-0.001 (0.000)**	-0.001 (0.000)**	-0.001 (0.000)**	-0.001 (0.000)***
Investment/GDP (in%)	0.002 (0.001)***	0.004 (0.001)***	0.003 (0.001)***	0.001 (0.001)	0.002 (0.001)***
Constant	0.015 (0.018)	-0.031 (0.040)	0.012 (0.019)	0.013 (0.037)	-0.039 (0.026)
Observations	385	366	366	366	366
R-squared	0.21	0.38		0.67	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 3: The effect of public consumption procyclicality on growth, Gaussian-weighted rolling window OLS method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government consumption)	-0.042 (0.019)**	-0.060 (0.023)***	0.000 (0.037)	-0.046 (0.020)**	-0.022 (0.033)
lag(Private credit/GDP)	-0.011 (0.004)**	-0.023 (0.007)***	-0.006 (0.011)	-0.024 (0.007)***	-0.028 (0.011)***
lag(Procyclicality of government consumption*Private credit/GDP)	0.043 (0.022)**	0.050 (0.024)**	-0.009 (0.042)	0.040 (0.021)*	0.015 (0.038)
lag(log (real GDP per capita))	-0.001 (0.001)*	-0.064 (0.018)***	-0.149 (0.021)***	-0.111 (0.022)***	-0.249 (0.030)***
Average years of schooling for the population over 25 years old	0.000 (0.001)	0.006 (0.004)	0.019 (0.004)***	0.004 (0.003)	0.010 (0.004)***
Openness	0.000 (0.000)*	0.001 (0.000)***	0.001 (0.000)***	0.000 (0.000)**	0.000 (0.000)
Inflation	-0.087 (0.029)***	-0.153 (0.036)***	-0.138 (0.030)***	-0.049 (0.031)	-0.070 (0.029)**
Population growth	-0.304 (0.295)	-1.038 (0.468)**	-0.678 (0.398)*	-0.647 (0.349)*	-0.292 (0.362)
Government share of GDP (in %)	0.000 (0.000)	-0.001 (0.000)**	-0.001 (0.000)**	0.000 (0.000)*	-0.001 (0.000)*
Investment/GDP (in%)	0.002 (0.000)***	0.003 (0.000)***	0.004 (0.000)***	0.002 (0.000)***	0.003 (0.001)***
Constant	-0.015 (0.017)	-0.566 (0.174)***	-1.443 (0.109)***	-0.903 (0.186)***	-2.024 (0.126)***
Observations	453	434	434	434	
R-squared	0.14	0.33		0.57	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.

Table 4: The effect of public investment procyclicality on growth, Gaussian-weighted rolling window OLS method

	OLS	Country f.e.		Country year f.e.	
		Standard	Corrected for AR(1) errors	Standard	Corrected for AR(1) errors
lag(Procyclicality of government investment)	-0.095 (0.036)***	-0.055 (0.039)	0.050 (0.086)	-0.064 (0.033)*	-0.058 (0.141)
lag(Private credit/GDP)	-0.007 (0.004)*	-0.012 (0.007)*	0.002 (0.012)	-0.016 (0.007)**	-0.015 (0.016)
lag(Procyclicality of government investment*Private credit/GDP)	0.098 (0.034)***	0.034 (0.032)	-0.102 (0.081)	0.029 (0.027)	-0.046 (0.125)
lag(log (real GDP per capita))	-0.002 (0.001)	-0.074 (0.018)***	-0.237 (0.024)***	-0.128 (0.025)***	-0.638 (0.041)***
Average years of schooling for the population over 25 years old	0.001 (0.001)	0.005 (0.004)	0.026 (0.004)***	0.002 (0.003)	0.005 (0.006)
Openness	0.000 (0.000)	0.001 (0.000)***	0.002 (0.000)***	0.000 (0.000)**	0.000 (0.000)
Inflation	-0.032 (0.020)	-0.096 (0.023)***	-0.123 (0.030)***	-0.034 (0.027)	-0.018 (0.026)
Population growth	-0.902 (0.321)***	-1.303 (0.585)**	-0.367 (0.449)	-0.685 (0.456)	0.201 (0.399)
Government share of GDP (in %)	0.000 (0.000)	0.000 (0.000)	-0.001 (0.000)*	0.000 (0.000)	0.000 (0.000)
Investment/GDP (in%)	0.001 (0.000)***	0.003 (0.001)***	0.005 (0.001)***	0.002 (0.000)***	0.006 (0.001)***
Constant	-0.008 (0.019)	-0.705 (0.176)***	-2.334 (0.095)***	-1.123 (0.216)***	-5.136 (0.063)***
Observations	453	434	434	434	434
R-squared	0.11	0.3		0.55	

Robust standard errors in parentheses
* significant at 10%; ** significant at 5%; *** significant at 1%

Note: The explained variable is the first difference of the log of real GDP per capita. In columns 3 and 5, we allowed for AR(1) autocorrelation in the error term by using Stata's command xtregar, which implements the method described in Batalgi (2001).

Source: OECD Economic Outlook, Levine dataset, Barro Lee dataset, Penn World Tables 6.1.