

Distortions in the Neoclassical Growth Model: A Cross-Country Analysis*

Pedro Brinca[†]

November 3, 2013

Abstract

This paper investigates the properties of distortions that manifest themselves as wedges in the equilibrium conditions of the neoclassical growth model across a sample of OECD countries for the 1970-2011 period. The quantitative relevance of each wedge and its robustness in generating fluctuations in macroeconomic aggregates is assessed. The efficiency wedge proves to be determinant in enabling models to replicate movements in output and investment, while the labor wedge is important to predict fluctuations in hours worked. Modeling distortions to the savings decision holds little quantitative or qualitative relevance. Also, investment seems to be the hardest aggregate to replicate, as prediction errors concerning output and hours worked are typically one order of magnitude smaller. These conclusions are statistically significant across the countries in the sample and are not limited to output drops. Finally, the geographical distance between countries and their degree of openness to trade are shown to contain information with regard to the wedges, stressing the importance of international mechanisms of transmission between distortions to the equilibrium conditions of the neoclassical growth model.

JEL Classification: E27, E30, E32, E37

Keywords: Business cycle accounting, frictions, economic fluctuations

*I am grateful for comments from Luis Aguiar-Conraria, Tobias Broer, Martin Flodén, John Hassler, Ellen McGrattan, Joana Pinho, Alexandra Ramos, Vincenzo Quadrini, Alper Çenesiz, IIES, CEFUP, and GEMF seminar participants, PET2013 and the 7th Annual meetings of the PEJ conference participants and, in particular, Sirius Dehdari for excellent research assistance.

[†]Correspondence to Pedro Brinca, email pedro.brinca@ne.su.se. Address: Department of Economics, Stockholm University, SE-106 91 Stockholm, Sweden.

1 Introduction

The focus of this paper concerns the measurement and systematic analysis of different types of distortions to the equilibrium conditions of the neoclassical growth model. Their quantitative relevance in generating fluctuations at the business cycle frequency in macroeconomic aggregates is also analyzed and tentative structural explanations for such distortions are put forth by identifying relevant indicators that contain information with respect to the said distortions.

The interest in analyzing the properties of deviations from theoretically postulated relationships among macroeconomic aggregates within the neoclassical framework can be traced back at least to Solow (1957). Deviations from observed output and capital and labor inputs for a given aggregate production function were taken to be the source of long term growth and became known as the 'Solow residual' or total factor productivity. Growth accounting exercises became widespread in order to measure the contribution of each factor with respect to changes in output.

This was mainly a growth issue until Kydland and Prescott (1982) introduced a multiplicative persistent shock into an aggregate production function and managed to generate fluctuations in macroeconomic aggregates at business cycle frequencies. By then this was done in a context of a general equilibrium model, with endogenous labor supply and savings decision. Subsequent work aimed to provide structural explanations for these shocks as well as creating departures from the neoclassical growth model that could replicate fluctuations observed in the data. However, much of the focus was still on total factor productivity and in theories that could explain it.

Researchers started to be interested in the properties of deviations in other equilibrium conditions, such as the labor-leisure choice. Mulligan (2002) looks into data for the U.S. from 1889 until 1996 to describe the statistical properties of such deviations (in this case to the labor-leisure choice) and provide tentative explanations behind them. Ohanian et al. (2008) perform a similar analysis to OECD countries. Other authors focused on models with financial frictions, such as Calstrom and Fuerst (1997) or Bernanke et al.

(1999), that express themselves mostly as distortions to the savings decision.

Much in the fashion of growth accounting, a business cycle accounting methodology was developed by Chari et al. (2007). Distortions to the equilibrium conditions, of what the authors dub a prototype growth model containing the key ingredients of the neoclassical framework, are measured. Their quantitative and qualitative relevance in generating fluctuations in macroeconomic aggregates through a series of simulations is assessed. In the work cited, the methodology is applied to both the Great Depression and the 1981 recession in the U.S.

Since then a large body of literature has developed based on Chari et al. (2007) methodology. Some authors provide methodological departures from Chari et al. (2007). Two examples can be found in Otsu (2009) that conducts the analysis in the context of a two country model and Sustek (2010), adding a Taylor type rule for nominal interest rate setting and government bonds.

Other have applied the methodology to other countries. Kobayashi and Inaba (2006) for Japan, Simonovska and Söderling (2008) for Chile and Lamas (2009) for Argentina, Mexico and Brazil are just few examples. The results seem to conclude, much in line with Chari et al. (2007), that total factor productivity and distortions to the labor choice are relevant, where distortions to the savings decision are considerably less important. Some authors focus their analysis to one type of deviations as in Restrepo-Echavarria and Cheremukhin (2010) or Cociuba and Ueberfeldt (2010) with distortions in the labor choice or other numerous studies concerning total factor productivity such as Islam et al. (2006). Finally, other line of work looks into a selected sample of countries and into specific periods of fluctuations such as output drops (see Dooyeon and Doblas-Madrid (2012) as one example).

This paper contributes to the literature in several dimensions. First the sample of countries chosen for analysis is driven purely by data availability. This avoids sample selection bias. Most business cycle accounting exercises restrict their samples by analyzing recessions, and, consequently, countries that experienced the recessive episodes. The validity of the conclusions are therefore restricted to the criteria that drove the sample selection.

Second, by taking a comparably large sample of countries and pooling the

measured distortions cross-sectionally when analyzing specific episodes that are perceived as having an international scope such as the oil shocks in the 1970s or the 2008 financial crisis, we can draw inference as to whether such episodes systematically generated distortions in the equilibrium conditions of the neoclassical growth model across the countries in the sample. In particular, an analysis of the period corresponding to the Great Recession allows us to provide robustness and statistical significance with regard to previous findings from Ohanian and Raffo (2012).

Third, by using data that goes back until 1970 at the quarterly frequency, we can decompose the distortions in their trend and cycle components. Mulligan (2002) highlights the importance of analyzing trend and cycle separately. The author finds that marginal tax rates are important in explaining the trend but not the cycle component of distortions to the labor decision. The relevance of taxes in explaining movements in the labor wedge is also highlighted by Ohanian et al. (2008), but, as before, only with respect to the trend.

Fourth, by applying the business cycle accounting methodology, we are able not only to measure and decompose the distortions but also to assess their quantitative relevance in generating fluctuations in macroeconomic aggregates that resemble movements in observed data. This was performed both for specific episodes and for the whole sample, and draw inference on whether specific distortions are systematically important across the countries. We compare simulations with observed data and determine the key distortions to be modeled in order to bring the neoclassical growth model closer to reality.

Lastly, these distortions are analyzed by country characteristics, in search for indicators that contain information with respect to the distortions and in this way suggest tentative extensions to the business cycle model that are general enough to be relevant for most countries in the sample. Two dimensions relevant for the trade literature, for example, are analyzed.

One is the geographical distance between the countries and the potential that may have with respect to the degree of comovement between the wedges across countries. To mind come iceberg costs, a standard way to model costly

trade, introduced by Samuelson (1954), and that became widespread in the trade literature.

The other is the degree of openness to trade and its comovement with each of the wedges. The effects of openness to trade with regard to the likelihood of a country experiencing a financial crisis has been addressed in the literature is one example of the relevance of this indicator. Cavallo and Frankel (2008) for example empirically show that the degree of trade openness decreases the vulnerability of countries to sudden stops.

2 Data, model and calibration

2.1 Data

The data used to measure the wedges comes from OECD Economic Outlook. It concerns GDP, Government Consumption, Gross Fixed Capital Formation, Imports, Exports and their respective deflators, Total Employment and Total Hours Worked per Employee. Additionally there is also data on total population and percentage population below 16 and percentage population over 65. All series end on the last quarter of 2011 and, in very few exceptions such as some series concerning hours worked and population, data ends in the last quarter of 2010 and is extrapolated to end in 2011Q4).

Based on this, data are transformed according to the following procedure. Deflators are transformed to have 2005Q1 as base year. Data is then deflated accordingly. The four observables are output, hours worked, investment and government consumption (plus net exports), all in per-capita units. Sales and indirect taxes are not taken into account in the computation of model output because of availability and comparability. Hence, output, investment and government consumption plus net exports per capita are just the deflated series divided by quarterly interpolated active population. Hours worked are the product of Total Employment and Hours Worked per Employee divided by active population.

2.2 Model

The prototype economy is the same as in Chari et al. (2007). It is the neo-classical growth model with labor and savings decisions and four exogenous random variables. These variables are the efficiency wedge A_t , the labor wedge τ_{lt} , the investment wedge τ_{xt} and the government wedge g_t . Consumers maximize expected utility over per-capita consumption c_t and labor l_t :

$$\max E_0 \sum_{t=0}^{\infty} \beta^t u(c_t, 1 - l_t) \quad (1)$$

subject to the budget constraint

$$c_t + (1 + \tau_{xt})x_t = (1 - \tau_{lt})w_t l_t + r_t k_t + T_t \quad (2)$$

and the capital accumulation law

$$(1 + \gamma_n)k_{t+1} = (1 - \delta)k_t + x_t \quad (3)$$

where x_t is investment, w_t the wage rate, r_t the rental rate on capital, β the discount factor, $1 + \gamma_n$ is the population growth rate, k_t the stock and T_t lump sum transfers, all in per capita terms. The production function is given by $A_t F(k_t, (1 + \gamma_z)^t l_t)$ where y_t is per capita output and γ_z the rate of labor augmenting technical progress. The representative firm maximizes profits and pays factors their marginal products. The equilibrium in the economy is therefore pinned down by the aggregate resource constraint

$$c_t + x_t + g_t = y_t \quad (4)$$

where y_t is per capita output, the production function

$$y_t = A_t F(k_t, (1 + \gamma)^t l_t) \quad (5)$$

the labor-leisure choice

$$-\frac{u_{lt}}{u_{ct}} = (1 - \tau_{lt})A_t(1 + \gamma)F_{lt} \quad (6)$$

and the savings optimality condition

$$u_{ct}(1 + \tau_{xt}) = \beta E_t[u_{c,t+1}(A_{t+1}F_{k,t+1} + (1 - \delta)(1 + \tau_{x,t+1}))] \quad (7)$$

where a function's subscript denotes the derivative of the function with respect to the subscript argument, evaluated at t . It is also assumed that g_t fluctuates around the trend $(1 + \gamma_z)^t$.

2.3 Functional forms and calibration

The utility function is additive separable in logarithmic consumption and leisure, i.e., $u(c, l) = \log(c) + \psi \log(1 - l)$. The production function is linear homogeneous in capital and labor i.e. $F(k, l) = k^\theta l^{(1-\theta)}$. The values used for the parametrization of the models are the ones taken by Chari et al. (2007), with the exception of the population growth rate which is country specific. The growth rate of labor-augmenting technical change is taken from Kehoe and Prescott (2007).

Country specific calibration of the parameters for each economy was not performed because I didn't want cross-country differences to be driven by different parametrization but rather by the distortions themselves. The values are shown in Table 1 below, at annualized rates:

Table 1: Calibration

γ	β	δ	ψ	θ
0.02	0.97	0.05	2.24	0.35

Given the values for the parameters in the table above, the model is solved for the steady-state quantities and the equilibrium is found. Equilibrium decision rules are derived assuming that the exogenous states (the wedges) follow a four dimensional vector auto-regressive of order one where the error process is assumed to be multivariate normal with mean zero and variance-covariance matrix $Q = B'B$ as described below:

$$\omega_{t+1} = P_0 + P\omega_t + \epsilon_{t+1}, \quad \epsilon \sim MVN(0, B'B) \quad (8)$$

The data is used as observables and the Kalman filter used to back out the innovations (wedges). The likelihood of the innovations being jointly normal is computed and the optimization program concerns the choice of the parameters of the VAR, i.e., the vector P_0 and the matrices P and B , such that the likelihood is maximized. This process is repeated for each country. Stationarity is imposed in the estimation.

The purpose of performing simulations is to see to what extent models with just one distortion or a combination of distortions have the ability to replicate observed data. Hence, new decision rules are computed, setting the wedges, that are excluded in a specific simulation exercise, to their unconditional mean values throughout the simulation procedure. Since in they no longer are random variables in the simulations, the computation of new equilibrium decision rules and allocations in the simulated economies ensures that agent's expectations are consistent with the model.

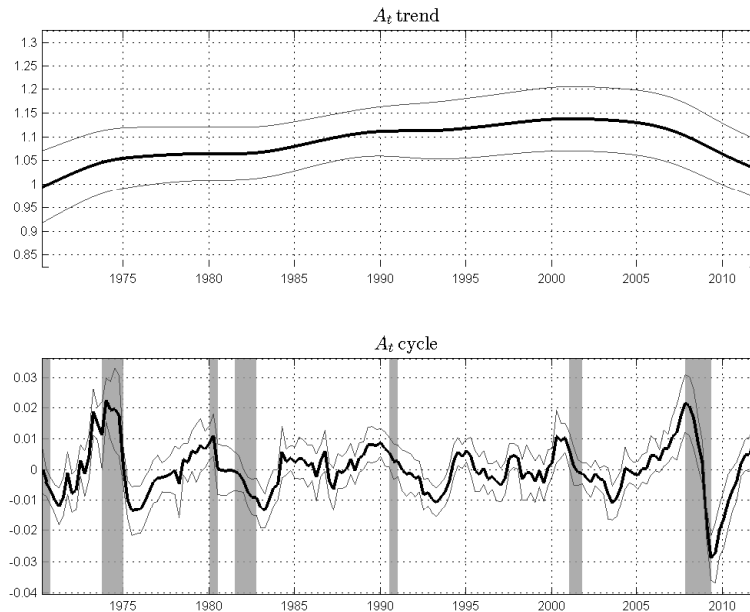
3 Wedges' trends and cycles

The wedges are filtered using the HP-filter as in Hodrick and Prescott (1997), with a smoothing factor of 1600. The original series, cycles and trends are presented in Appendix A. Average trends and cycles are computed by taking cross-country per quarter averages. Confidence intervals for the average trend and cycles are computed by drawing with replacement sample trend and cycles and computing their average. The empirical distribution of the average components is then used to compute the confidence intervals at the desired significance level. The shaded quarters in Figures 1 to 4 and 6, correspond to periods for which the NBER declared the U.S. economy to be in recession. The relevance of choosing the U.S. cycle as reference comes from a body of literature that investigates the relevance of the U.S. cycle for the business cycles of other economies. In particular Kose et al. (2012), analyze the issue using a positive comovement between the U.S. cycle and the common component of the rest of the world's business cycles as identification assumption.

3.1 The efficiency wedge

The average trend for the efficiency wedge shows a modest positive slope until the early 2000's, and since then a steeper decline. For most of the sample (1975-2010), average detrended total factor productivity is significantly above one, indicating that its contribution for growth has been on average above the 2% that Kehoe and Prescott (2007) use. There is a slowdown after this period that was aggravated at the early stages of the 2008 financial crisis. The confidence intervals suggest that the series is relatively homoscedastic at the cross-sectional dimension as the amplitude remains fairly constant over the sample period.

Figure 1: Average trend and cycle for the efficiency wedge



95% Confidence intervals computed with bootstrapping, 1000 draws
Shaded areas indicate U.S. recessions as declared by the NBER

With respect to the cyclical component we can see that fluctuations in total factor productivity are remarkably synchronized in the sense that for many periods the fluctuations are significantly different from zero. The most

notable periods of accentuated fluctuations coincide with the Yom Kippur war and the oil crisis that ensued the autumn of 1973 and the 2008 financial crisis. It is notable however that the periods the NBER declared to mark recessions in the U.S. economy coincide with the extreme realizations of the wedges, given that the data was aggregated giving equal weight to each of the 19 countries in the sample. This suggests the weight that the U.S. economy still carries in determining business cycles for the countries in the sample.

3.2 The labor wedge

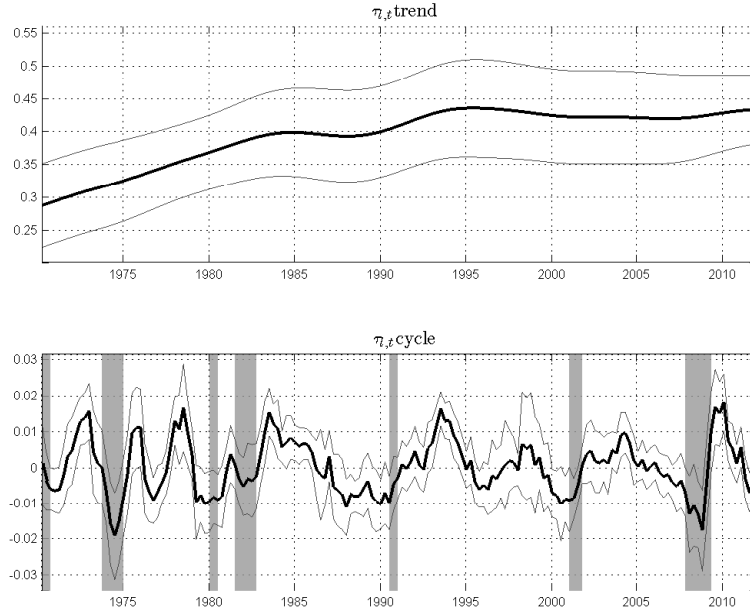
In Figure 2 we see that there is an overall tendency for the labor wedge to increase over the sample period and the increase is statistically significant. The average trend rose from just below 0.30 to 0.45. As in the case for the efficiency wedge, the cross-sectional variance is fairly stable. Nonetheless there are some countries, namely the U.S. and Canada, for which the trend is downward slopping. This is also documented by Shimer (2010) and Cociuba and Ueberfedt (2010) for the U.S. economy.

Concerning the cyclical component of the labor wedge, we can see that it is also fairly synchronized over the sample period. In fact there are, as in the case for total factor productivity, many instances where the aggregate labor wedge is statistically different from zero. The cycle is also similar in amplitude to the efficiency wedge, i.e., fluctuations are of about $\pm 3\%$. The series in itself, however, is more volatile, with more episodes comparable in magnitude to the fluctuations observed in the early 1970s and late 2000s.

As in the case of the efficiency wedge, the larger deviations from trend coincide with U.S. recessions, though where the efficiency wedge peaks at those, here the recessions coincide with periods for which the labor wedge is at its lowest. With regard to the U.S., Shimer (2009) finds that the cyclical component of the labor wedge rises during recessive periods. In the case for the efficiency wedge, the drop in the sample average was coincident with U.S. recessions, but in this case there seems to be a lagging effect. As can be observed in Figure 2, recessions in the U.S. coincide with local minima of the average labor wedge in our sample, and it indeed rises during, or shortly

after, later periods of the recessions.

Figure 2: Average trend and cycle for the labor wedge



95% Confidence intervals computed with bootstrapping, 1000 draws
Shaded areas indicate U.S. recessions as declared by the NBER

Shimer (2009) argues that two obvious explanations would be that labor and consumption taxes rise during recessions.

Note that in our prototype economy there are no consumption taxes but these would be captured by the labor wedge. To see this, notice that if the budget constraint in (2) would include taxes on consumption:

$$(1 + \tau_{ct})c_t + (1 + \tau_{xt})x_t = (1 - \tau_{lt})w_t l_t + r_t k_t + T_t \quad (9)$$

the labor leisure choice would then be:

$$-\frac{u_{lt}}{u_{ct}} = \frac{1 - \tau_{lt}}{1 + \tau_{ct}} A_t (1 + \gamma) F_{lt} \quad (10)$$

Since the labor wedge is computed residually to make the marginal rate of

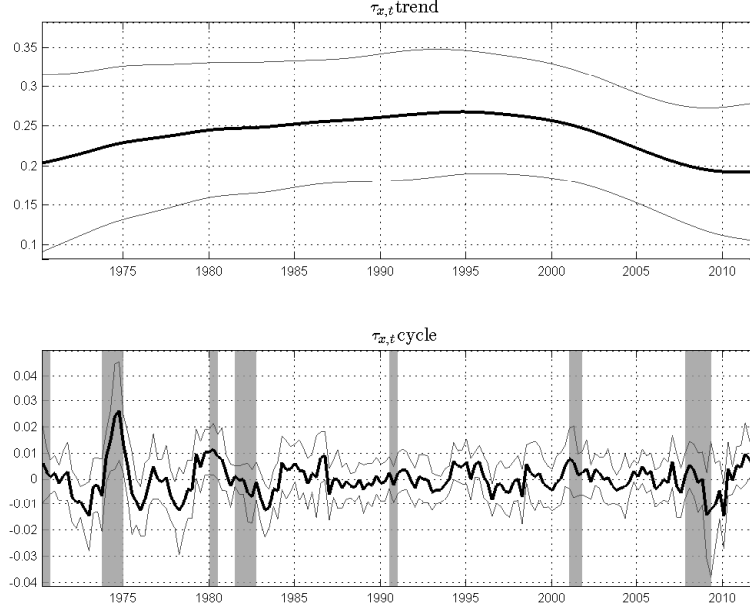
substitution between labor and consumption to equate the marginal product of labor, the labor wedge reflects changes in $\frac{1-\tau_{lt}}{1+\tau_{ct}}$. Shimer (2009) cites McGrattan and Prescott (2009) in arguing that changes in consumption taxes fit the data much better than tax changes in labor income. Mertens and Ravn (2008) however, put an upper bound of 18% to the variance of output explainable by tax shocks at the business cycle frequency. As argued before, this underlines the importance of decomposing the labor wedge between trend and business cycle frequencies.

3.3 Investment wedge

The investment wedge, unlike with the previous two cases, exhibits much larger cross-sectional volatility. Though the point estimates in Figure 3 show a slight rise of the investment wedge until the late 1990s and a subsequent decline until the end of sample, the cross-sectional variance is such that we cannot reject that the average investment wedge was constant throughout our sample.

With respect to the cyclical component, both the amplitude of the deviations and the volatility are higher than in the previous two cases. Also, we find much fewer instances with statistical significance for average cyclical movements. This suggests that there is little synchronization in the sample with regard to distortions to the savings decision. Most notably, the period with the largest deviation from the trend was in the early 1970s and, unlike in the previous cases, there is hardly any co-movement with regard to the investment wedge for the last financial crisis. If we restrict ourselves though to the analysis of the point estimates, we can still partially observe the previous pattern of the wedges peaking during U.S. recessions, namely during the mentioned 1970's period, the 1980's slowdown and the last financial crisis.

Figure 3: Average trend and cycle for the investment wedge



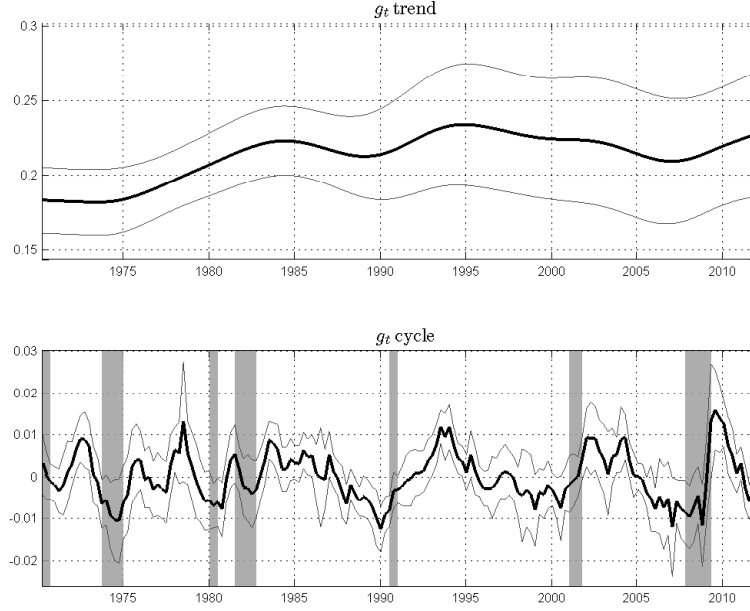
95% Confidence intervals computed with bootstrapping, 1000 draws
Shaded areas indicate U.S. Recessions as declared by the NBER

3.4 Government wedge

The government wedge (government consumption plus net exports) is expressed as a fraction of output. The trend is not nearly as smooth and there is an increasing dispersion though there is a marginally significant increase in the trend component of the government wedge over the sample period.

With regard to the cyclical variation, as in the case for the efficiency and labor wedges, there are many instances where the average cycle is comparable in magnitude and also statistically different from zero. U.S. recessions are still a good indicator of local minima for the government wedge, though as before, there are many other instances where the deviations are statistically significant.

Figure 4: Average trend and cycle for the government wedge



95% Confidence intervals computed with bootstrapping, 1000 draws
Shaded areas indicate U.S. recessions as declared by the NBER

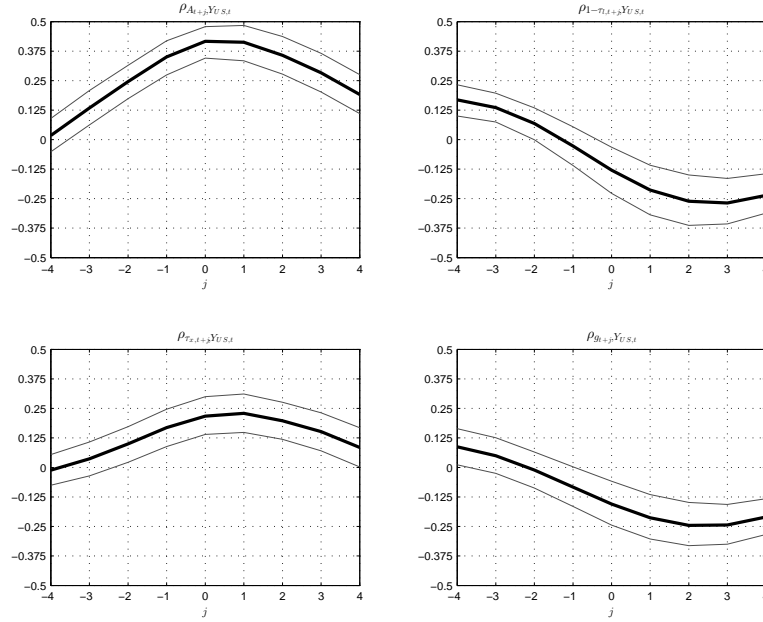
3.5 Summary statistics

We have seen that movements in the efficiency, labor and government wedge are fairly synchronized and that have had changes in the trend component that are statistically different during several periods in the sample. The same cannot be said about the investment wedge. This indicates that there is a greater disparity in shocks to the savings decision than it is the case for the other equilibrium conditions, given the much fewer instances in which deviations from trend for the investment wedge behaved similarly enough such that they were significantly bigger(smaller) than zero.

We saw that U.S. recessions contain information regarding some features of the wedges. For example, U.S. recessions seem to lead significant drops in TFP and lead significant increases in the labor wedge. The Figure 5 below shows the lead-lag cross-correlation structure between each of the average

wedges' cycles and the U.S. output cycle.

Figure 5: Lead-lag cross-correlation between ω_{t+j} and $Y_{US,t}$



95% Confidence intervals computed with bootstrapping, 1000 draws

The U.S. cycle is positively correlated with TFP, with the higher point estimates suggesting it to be a coincident or leading (by one period) indicator of the rest of the sample average. With respect to the labor wedge, the contemporaneous correlation is negative, a result similar to what Shimer (2009) finds for the U.S. economy, and the estimate is statistically significant. The highest point estimate of the correlation (in absolute value) occurs with a three quarters lag making the U.S. cycle a leading indicator for the average labor wedge. The investment wedge shows a similar correlation structure as the efficiency wedge i.e. the correlation is positive and the U.S. output cycle is a coincident or leading indicator. In the case of the government wedge, the highest point estimate for the average correlation is obtained for $j = 3$ and it is negative. Note that in all cases, the correlations are 'skewed to the right' i.e. the higher point estimates are mostly found for $j \geq 0$. This provides

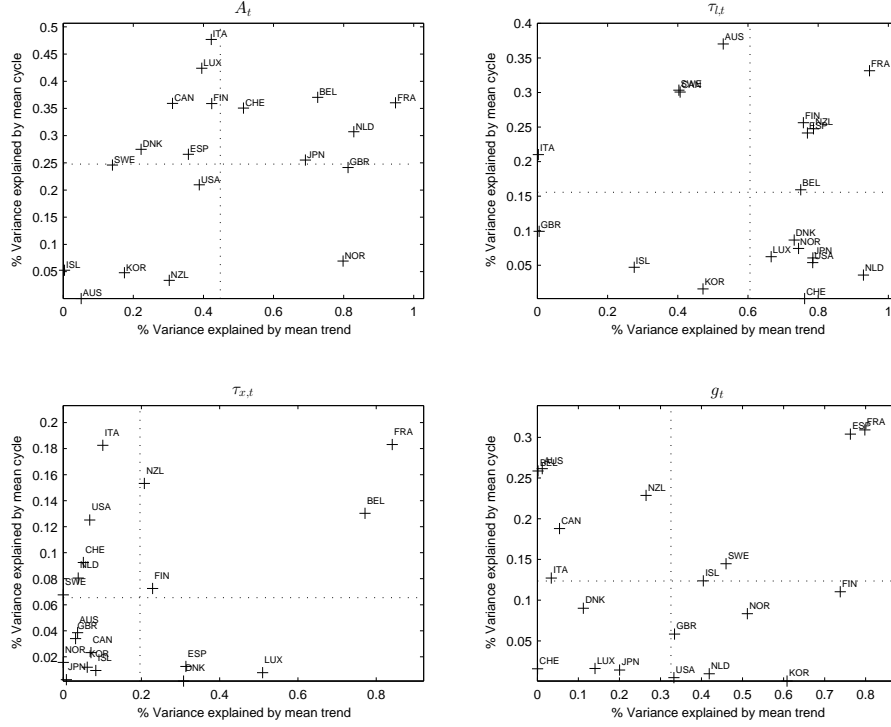
further evidence of the relevance that the U.S. cycle may have with regard to the wedges in the rest of the countries in our sample.

In Figure 6, the percentage variation explained by both the mean cycle and variance can be depicted, for each country in the sample. The total variance explained is obtained by regressing the individual series on the average components (minus the respective series) and reporting the R^2 's of the regressions. Mean trend and cycle of the investment wedge explain less of the variation in the individual series compared to the other wedges. The average trend explains around 20% of the individual trends, against 45% for the efficiency wedge, 61% for the labor wedge and 33% of the government wedge. For the cycle, the differences are similar, with only 6% of the investment wedge cycle being explained by the average cycle against 25%, 15% and 13% for the efficiency, labor and government wedges respectively. This confirms our results that the investment wedge is significantly less synchronized than the other wedges, for both trend and cycle.

Another interesting aspect of Figure 6 is how France is the country whose wedges are most correlated with the average trend throughout the whole sample. With regard with the common cyclical component, France's wedges are also along the ones which show a higher degree of synchronization. A possible explanation is that out of our sample, seven countries are part of the Euro Area. Following Aguiar-Conraria and Soares (2011), France and Germany were found to be the core of the Euro Area i.e. the most synchronized countries with the rest of Europe. Germany is not part of the sample due to data issues¹, but the fact that France is so synchronized with the average components of the wedges lends support to previous findings from Aguiar-Conraria and Soares (2011).

¹No available data on hours worked for Germany before the reunification

Figure 6: Percentage variation explained by average trend and cycle



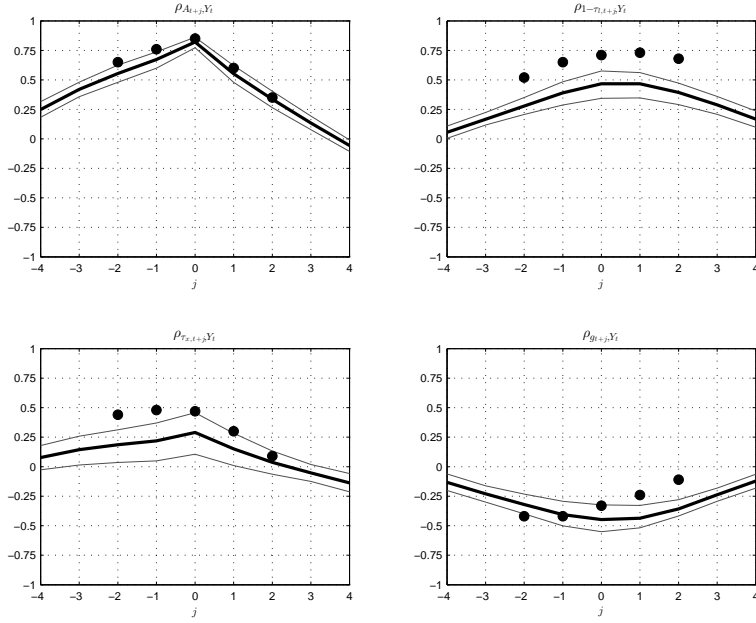
Vertical and horizontal lines concern total variance across all series explained by average cycle and trend respectively

Figure 7 shows the cross-correlations between each of the (HP-filtered) wedges and cyclical output. The patterns of correlation between each of the wedges and cyclical output is similar enough across countries such that it allows us to draw statistical significance. The first observation is that the wedges are coincident indicators of cyclical output i.e. the absolute value of the cross-correlation reaches its highest value for the contemporaneous correlation.

The efficiency and investment wedges are procyclical, even though the persistence and magnitude of the procyclicality is higher in the case of the efficiency wedge. These results are in line with Chari et al. (2007). The counter-cyclicality of the government wedge is also in line with Chari et al. (2007) but not for the case of the labor wedge. In fact, for the countries and

periods in the sample (and even for the U.S.), the labor wedge is counter-cyclical.

Figure 7: Lead-lag cross-correlation between ω_t and Y_{t+j}



95% Confidence intervals computed with bootstrapping, 1000 draws
 • - Estimates from Chari et al. (2007), for the U.S. economy

The cross-correlation structures found are similar enough across countries such that inference can be drawn at the 5% significance level, for all contemporaneous correlations and at least for one lead and lag. Comparing the results with Chari et al. (2007) with respect to the efficiency wedge, the similarity is striking. Point estimates reported in Chari et al. (2007) also show the efficiency wedge to be a coincident indicator for output. Their point estimate, 0.85, is very similar to ours, 0.82, and lies within our estimate's 95% confidence interval of $[0.77, 0.87]$. With respect to the cross-correlation at different lags, we find our estimates to be of the same order of magnitude and for $j = 1, 2$ we can reject that they are statistically different.

When it comes to the labor wedge, our results are statistically smaller in magnitude but qualitatively in line with Chari et al. (2007). We find the

average labor wedge to be procyclical on average in our sample, and this to be statistically significant for all 4 lags/leads. The only exception, i.e. a countercyclical labor wedge is the Republic of Korea. Our findings are also in line with more recent work from Shimer (2009)².

In the case of the investment wedge, we find it to be moderately procyclical, on average, smaller in magnitude than what Chari et al. (2007) find for the U.S. Finally, the government wedge is found to be countercyclical, though in this case our findings are much closer to what Chari et al. (2007) find for the U.S.

The similar behavior of the cross-correlations of each wedge with output in Figure 7 revives Lucas' idea that all business cycles are alike, though the evidence with respect to the investment wedge is somewhat weaker.

Another feature studied in Chari et al. (2007) concerns the relative volatility of the wedges to output. Table 2 below, presents point estimates for the average relative volatility for each of the wedges and the associated 95% confidence interval.

Table 2: Standard Deviation Relative to Output

A_t	$\tau_{l,t}$	$\tau_{x,t}$	g_t
0.89	1.00	1.11	0.82
(0.81,0.97)	(0.89,1.14)	(0.87,1.37)	(0.69,0.99)

95% Confidence intervals computed with bootstrapping, 1000 draws

In our findings, the standard deviation of the efficiency wedge relative to that of output is found to be statistically higher than the one Chari et al. (2007) find for the U.S. (0.63). Our findings differ also in terms of the government wedge, which in our case is significantly smaller than the one found by the same authors (1.51). However, in the case of the labor and investment wedges, Chari et al. (2007) estimates of their relative volatility (of 0.92 and

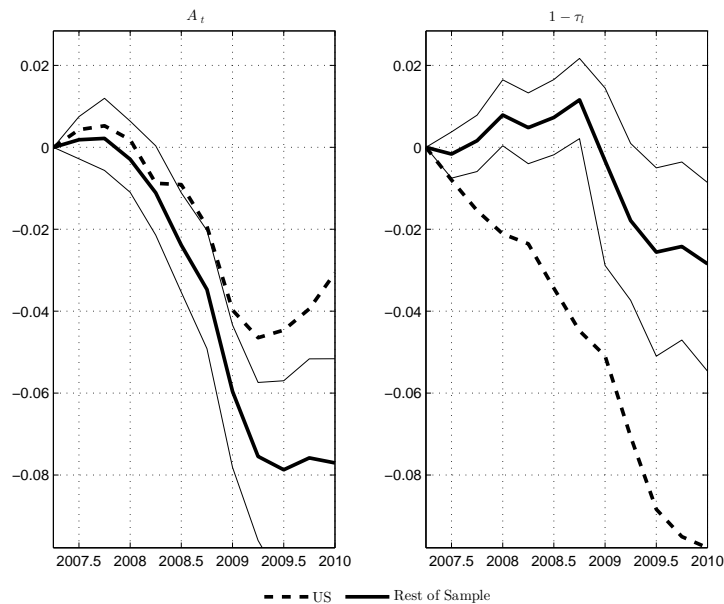
²Notice however that Shimer (2009) results differ from ours only in the sense that the definition of labor wedge is different. We follow Chari et al. (2007) in defining the labor wedge in Figure 7 as $1 - \tau_{l,t}$ for comparability. Everywhere else in the paper, the labor wedge is defined as in Shimer (2009), i.e., just $\tau_{l,t}$. As a consequence, Shimer (2009) obviously documents the labor wedge to be countercyclical instead.

1.18 respectively), fall within the 95% confidence interval computed for our estimates as shown in Table 2.

3.6 Efficiency and labor wedges in the Great Recession

In recent work by Ohanian and Raffo (2012), the authors look into total factor productivity and the labor wedge for the 2007-09 period. They conclude that for the U.S., the drop in the labor wedge is much higher than in total factor productivity. With respect to the other countries in their sample, the opposite happens: the Great Recession seems to be a total factor productivity phenomenon.

Figure 8: Efficiency and labor wedges in the Great Recession



95% Confidence intervals computed with bootstrapping, 1000 draws

The results in this paper are consistent with this finding, as shown in Figure 8. The wedges are normalized to their value in the first quarter of 2007 and confidence bands are computed for the averages of the efficiency and labor wedges in the sample (excluding the U.S.). The exception to this

rule are Ireland, Japan, Spain and Korea. Ireland is not included in the present analysis but the results with regard to the other countries are similar with the exception of Japan.

Quantitatively the results are somewhat different. The magnitude of the TFP drop for the U.S. is smaller in Ohanian and Raffo (2012) and the magnitude of the labor wedge is bigger. The authors use an accounting method with perfect foresight with regard to the future path of the wedges, instead of modeling expectations as in Chari et al. (2007). There are also differences in parametrization. The magnitude of the labor wedge for example, is directly affected by the time-preference parameter, as Chari et al. (2007) show. However, from a qualitative perspective, the results are almost identical and contribute to the literature by assessing the robustness and statistical significance of previous findings by Ohanian and Raffo (2012).

The fact that the Great Recession is experienced in such different ways between the U.S. and the other countries in the sample when seen through the lens of the prototype economy, casts doubts upon the view that the crisis was the outcome of similar banking crisis operating through the same channels, as Ohanian and Raffo (2012) argue. The authors stress the correlation between the relevance of the labor wedge and large movements in housing prices, observed in the U.S., Ireland and Spain as indicative that theories relating the labor wedge with the housing sector may prove fruitful.

Another possibly complementary view, is that crisis not perceived as structural may lead firms to hoard labor, to avoid firing/hiring costs for example, in the expectation that the economy will recover relatively faster. On the other hand, if the crisis is perceived as structural and leading to profound changes in the economy, firms will not hold on to unproductive workers. This could explain the pattern referred above i.e. for the U.S. where the crisis started and other countries that had large structural adjustments, the labor wedge plays a much larger role, and for countries where the crisis may have been perceived as a temporary phenomenon that originated in another country, firms hold on to their workers making labor productivity fall, with this showing up as a total factor productivity drop.

As an example, Sweden experienced two large recessions in the last 25

years: the burst of a financial bubble in the early 1990s and the effects of the Great Recession in the late 2000s. Brinca (2013) performs a business cycle accounting exercise to Swedish data for those periods and finds the labor wedge to play a much bigger role in the 1990s recession than for the 2007-2009 period, where total a total factor productivity decline explains most of the drop in output.

The author also finds that in the 1990s, labor productivity did not fall but unemployment increased a lot. During the recession in the late 2000s, labor productivity experienced a large drop but unemployment did not rise as much as previously and the economy also recovered much faster, consistent with the view that firms perceived the Great Recession period in Sweden as a temporary phenomena and hoarded labor, which led to the observed drop in labor productivity and the relevance of the efficiency wedge.

4 Simulations

Once measured the distortions, these can be considered the first best, with respect to the quantitative behavior that extensions that express themselves as distortions to the equilibrium conditions of the neoclassical growth model must exhibit. However, if the aim is to replicate movements observed in the macroeconomic aggregates, one needs also to assess the potential that the said distortions have to generate fluctuations in the data.

If we would simulate the model and feed the realizations of the four wedges as shocks i.e. the measured distortions, we would recover the original data. There is no surprise in this, since the distortions were measured precisely to make the equilibrium conditions hold with equality. However, if we do not feed all the measured wedges as shocks and simulate the model, in general, equilibrium allocations in the model and observed data will differ. The relevant question is then, by how much? If, for example, we model total factor productivity in such a way that we are able to exactly reproduce the efficiency wedge, how would equilibrium allocations compare with the data? Or, in general, if we would be able to model all but one distortion in a way that would replicate exactly the associated measured wedge, how far could

we go? An alternative way to assess the importance of a wedge is to pose the question of, how good is a model in replicating the data when all wedges but one, are included in the analysis.

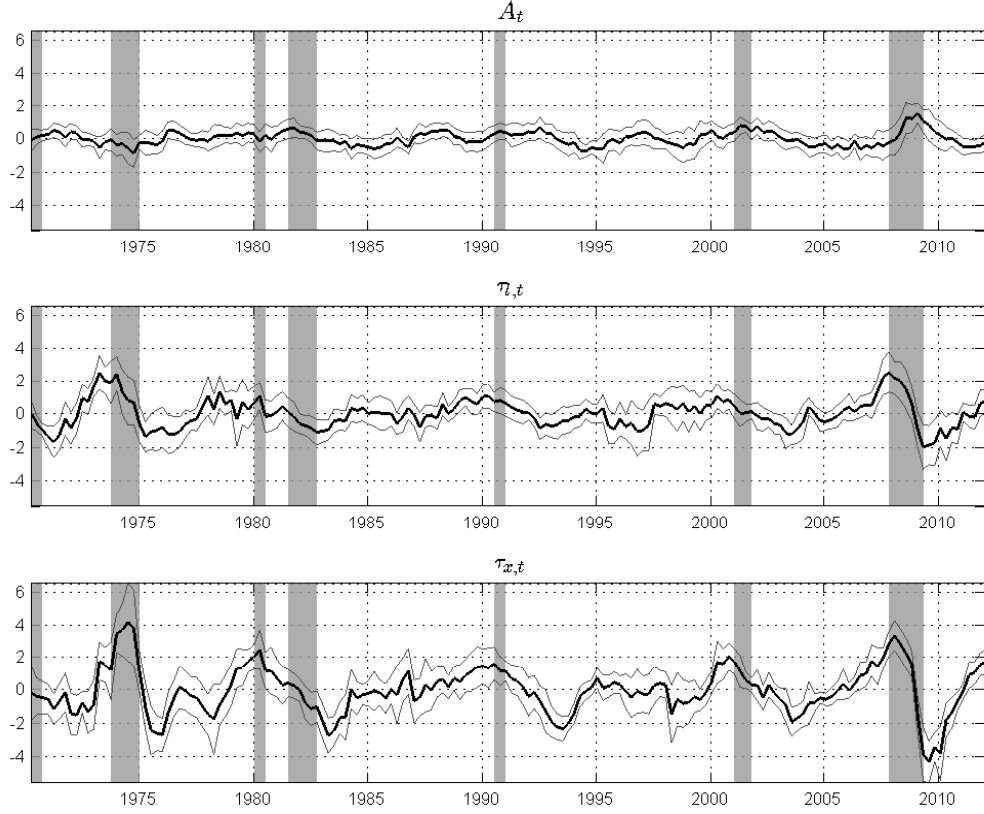
These questions are typical applications of business cycle accounting exercises. In Chari et al. (2007), evidence points towards the efficiency and labor wedge being key margins to be modeled in order to be able to replicate movements in output, hours and investment such as the ones observed in the data for the 1981 recession and the Great Depression of 1929. Most studies seem to converge to the same conclusion. This section also adds to the literature by checking the robustness of this common finding i.e. that modeling the efficiency and labor wedges contribute to a superior performance versus models that aim to replicate distortions in the savings decision.

In this section we assess the performance of models with just one wedge. The results regarding the analysis of models where all but one wedge are feed back to the model are included in the Appendix. For each country in the sample, I simulate the four observables and measure the deviations from observed data. In Figure 9, we plot simulations errors with regard to output. The difference between model prediction and realized output is pooled across the countries for each quarter and plotted against time. The results with regard to hours and investment are included in the Appendix.

The analysis of Figure 9 leads to several conclusions. First, economies with just the efficiency wedge have much fewer periods where the difference between observed and simulated output is similar enough across countries such that it is statistically different from zero. This only happens for 28 periods, against a total of 85 and 101 for the labor and investment wedge economies respectively.

Second, the efficiency wedge economies seem to produce the smaller deviations from the data, on average, followed by the labor and investment wedge economies. Third, the quarters for which simulated data more severely underestimate the magnitude of output deviations from trend coincide with recessions in the U.S. as declared by the NBER. This is even more remarkable given that the data in Figure 9 concerns (unweighted) aggregate data for the OECD countries.

Figure 9: Simulation Errors for Cyclical Output in One Wedge Economies



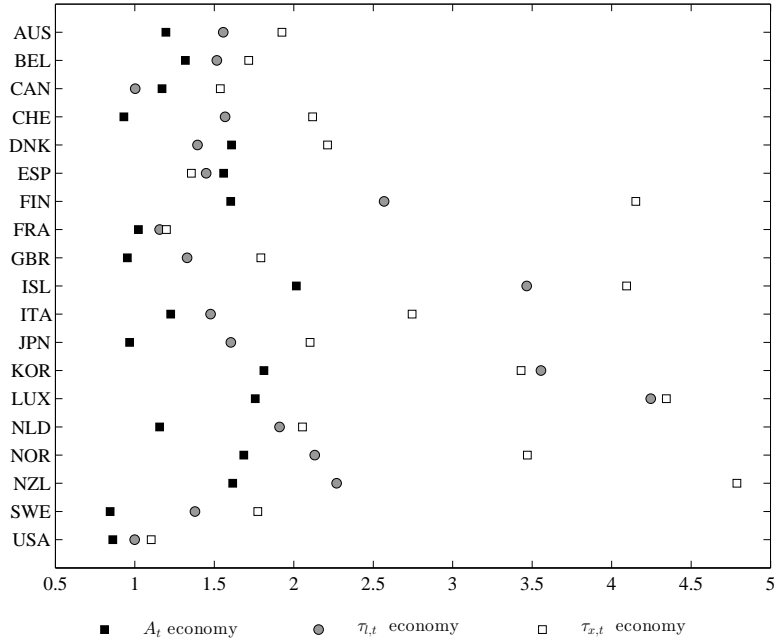
95% Confidence intervals computed with bootstrapping, 1000 draws
Shaded quarters indicate US recessions as declared by the NBER

It seems that all models systematically underestimate the magnitude of such recessions, even though this effect is more severe in economies without the efficiency wedge. In fact, in this case, only during the 2001 and 2008 recessions, simulation errors in the efficiency wedge economy have shown to be statistically significant, where for the other three economies, this happened for all the recessions in the sample period.

Note however, that since we are working with pooled data, smaller average deviations could just mean that the cross-sectional distribution is more symmetric, not that the errors are necessarily smaller. In order to check for that, Figure 10 shows the root mean square errors (RMSEs) for the three

types of economies, in the case of deviations from output cycle. The efficiency wedge economies produce the smallest RMSEs, followed by the labor wedge and then the investment wedge economies. This ordering is observed for almost all of the countries. Also average output RMSEs are 1.33 for the efficiency wedge, 1.92 for the labor wedge and 2.52 for the investment wedge economies.

Figure 10: RMSE's for Deviations from Output in One Wedge Economies



In order to have a statistical assessment of the comparative performance of the three types of models, Table 3 shows the outcome of parametric and non-parametric, joint and pairwise tests concerning the RMSEs produced. The difference between the average RMSEs across the three models is statistically significant at least at the 10% level for all comparisons.

The joint ANOVA test and its non-parametric equivalent Friedman's test both lead us to reject the null hypothesis that the average (median ranking of the) RMSEs for the three types of economies are equal.

In the first case, normality of the distribution of RMSEs is assumed and therefore the test statistic follows the F distribution. In the case of Fried-

man's test, normality is not assumed and only the ranking of the measured RMSEs is compared between the one wedge economies, across countries. In this case, Friedman's F statistic is asymptotically χ^2 distributed with two degrees of freedom.

Table 3: Statistical Tests of Comparative Performance

Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01	<i>t</i> -tests	A_t vs $\tau_{l,t}$	-2.62	0.01
Friedman	< 0.01		A_t vs $\tau_{x,t}$	-4.27	< 0.01
			$\tau_{l,t}$ vs $\tau_{x,t}$	-1.76	0.09
Economy	RMSEs	Wilcoxon			
A_t	1.33		A_t vs $\tau_{l,t}$	300.00	0.04
$\tau_{l,t}$	1.92		A_t vs $\tau_{x,t}$	237.00	< 0.01
$\tau_{x,t}$	2.52		$\tau_{l,t}$ vs $\tau_{x,t}$	308.00	0.07

In the case of the pairwise tests, *t*-tests were performed where the assumption is, again, that the average difference between the RMSEs is normally distributed. When using Wilcoxon's rank sum test we relax that assumption and take only the relative ranks into account. As in the case of the joint tests, there is no qualitative difference between the parametric and non-parametric testing if we are set to reject the null hypothesis of equal average (median ranking of the) RMSEs between the three types of economies at the 10% significance level. The finding is that, when it comes to output, the efficiency wedge is the key margin to be modeled, followed by the labor wedge and last the investment wedge.

As in Chari et al. (2007), in order to test the robustness of this finding, simulations are performed where all but one wedge are included. The equivalent to Figure 10 is reproduced in Appendix B and so is the equivalent to Table 3. The average of RMSEs for the economies with all but one wedge are of 2.04, 1.93 and 1.39.

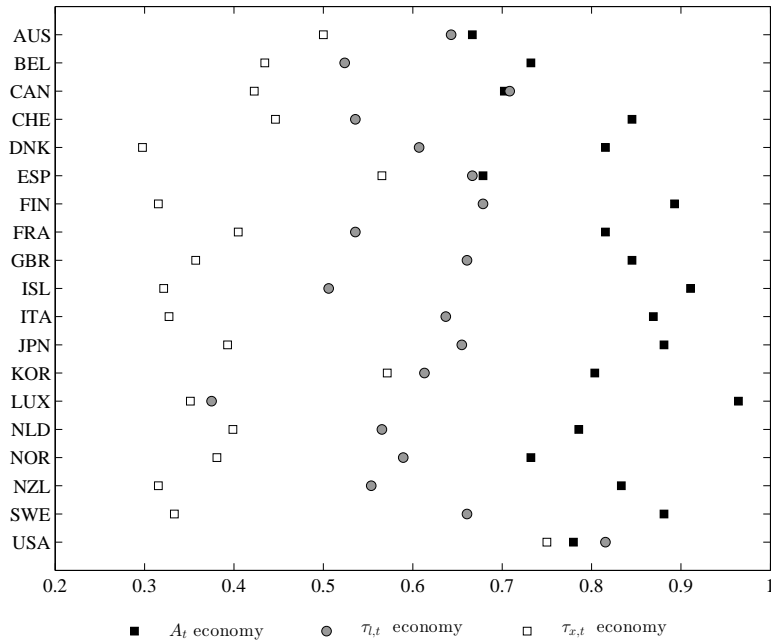
The economies with no efficiency wedge are the ones that perform the worst on average, followed by the economies with no labor wedge and lastly by the economies with no investment wedge, though only the comparisons of the no efficiency and no labor wedge economies against the no investment

wedge economies have statistical significance.

A simple analysis of the magnitude of the RMSEs can miss an aspect that might be relevant to the researcher. A model may produce smaller RMSEs but still lead to predictions that, on average, often are more qualitatively wrong than a model that tends to produce larger RMSEs but leads to predictions that are qualitatively correct i.e. predictions that correctly indicate an expansion or contraction of output in this case.

To control for that, in Figure 11 I show the success ratios for each country's output predictions of each type of the three economies mentioned before.

Figure 11: Success Ratios for Deviations from Output in 1 Wedge Economies



The statistic in this case indicates the percentage of times that simulated and observed output are of the same sign i.e. that simulated data is below/above trend when observed data is also below/above trend.

Figure 11 is even starker relative to previous findings. The efficiency wedge economies produce, on average, qualitatively correct predictions about 81% of the times, against 61% for the labor wedge and only 42% for the investment wedge economies.

It is worth noticing that just as the efficiency wedge economy produces RMSEs that are about half the size of the RMSEs associated with the investment wedge economies, the average success ratio for the efficiency wedge economy is about twice as high when compared to the same statistic regarding the investment wedge economies.

As before, statistical tests are performed and these differences are significant at the 1% level. The results also hold for all but one wedge economies, i.e., economies without the efficiency wedge have a success ratio of 51%, against 73% for the no labor and 87% for the no investment wedge economies.

Table 4: Statistical Tests of Comparative Performance

Output					
Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01	<i>t</i> -tests	A_t vs $\tau_{l,t}$	7.19	< 0.01
Friedman	< 0.01		A_t vs $\tau_{x,t}$	12.34	< 0.01
			$\tau_{l,t}$ vs $\tau_{x,t}$	5.66	< 0.01
Economy	SRs	Wilcoxon			
A_t	81%		A_t vs $\tau_{l,t}$	537.00	< 0.01
$\tau_{l,t}$	61%		A_t vs $\tau_{x,t}$	546.00	< 0.01
$\tau_{x,t}$	42%		$\tau_{l,t}$ vs $\tau_{x,t}$	511.50	< 0.01

All this leads to the conclusion that, in terms of quantitative and qualitative relevance, the efficiency wedge is the key margin to address fluctuations in output. It is worth noting that modeling only the efficiency wedge leads to overall smaller average RMSEs than including all other wedges and leaving out the efficiency wedge, i.e., 1.33 vs 2.04, and this result is even more clear if the success ratio is taken as measure: 81% vs 51%.

The labor wedge also plays a role as it leads to better predictions concerning output than the investment wedge. This result is robust to the performance measure used (the RMSEs - 1.92 vs 2.52 or the success ratios - 61% vs 42%) and it is statistically significant at the 10% level in the case of the labor wedge vs investment wedge economy (*t*-statistic of -1.76 with a *p*-value of 0.09 and Wilcoxon's *p*-value of 0.07). In the case of the no

labor wedge vs no investment wedge economy, the t -statistic is not significant (p -value of 0.12) but the Wilcoxon ranksum test leads us to reject the null hypothesis at the 5% significance level. The different conclusions in the joint tests suggests that even though the ordering for each type of economies' performance (ranks) shows a statistically significant pattern (p -value < 0.01 for Friedman's test), the magnitude of the differences between each type of economy for each country is quite heterogeneous, a result that can be confirmed by visual inspection of Figure 10 in Appendix B.

The investment wedge seems to be the less important margin to be modeled. It leads to higher RMSEs and lower success ratios. Also, unlike the previous two cases, economies where only the investment wedge is modeled do not even lead to predictions that are qualitatively correct 50% of the time. The average success ratio of the investment wedge economies is only of 42% and the null hypothesis of it being equal to 50% can be rejected (t -statistic of -3.24 with p -value < 0.01).

4.1 Hours and Investment

The above results establish the efficiency and labor wedge as key margins to be modeled in order to replicate fluctuations in output. However, the researcher may be interested in how the different models perform with regard to replicating fluctuations in the other two observables i.e. hours worked and investment. The figures and tables shown above for output are included in Appendix B also with regard to hours worked and investment.

With regard to hours worked, the point estimates for the mean RMSEs (1.66, 1.91 and 2.33 for the efficiency, labor and investment wedge economies respectively) still single out the efficiency wedge economies in producing the smaller RMSEs on average, though there is only statistical significance when it comes to comparing the efficiency to the investment wedge economy (t -statistic of -2.45 with a p -value of 0.02). When we look at the all-but-one-wedge economies, the results are different though. The no labor wedge economy is the one that produces the higher RMSEs, followed by the investment wedge and then the no efficiency wedge (1.91, 2.33 and

1.66 , respectively) and in this case, statistical significance is found for all the t -tests and all but one of the Wilcoxon ranksum tests. The results are shown in Appendix B, Table 2.

The labor wedge thus seems to be relatively more relevant when it comes to simulating data that resembles observed fluctuations in hours worked. This importance is even clearer if we look at the success ratios. The labor wedge economy predictions are qualitatively correct about 76% of the time, against 52% and 48% for the efficiency and investment wedge economies. The labor wedge seems key in order to make the model qualitatively in line with the data. This result is statistically significant for both comparisons of the success ratios of the labor wedge with the efficiency and investment wedge economies. The results are also confirmed by the analysis of the all-but-one-wedge economies, i.e., the no labor wedge economies fare comparably worse and the differences are statistically significant.

Lastly, with regard to investment, point estimates suggest the labor wedge again as being the most relevant with respect to producing the lower RMSEs on average (7.47, 9.74 and 11.80 for the labor, efficiency and investment wedge economies respectively). The all-but-one wedge economies also point in the same direction, though in both cases, there is no statistical significance. However, in terms of the success ratios, as it was the case for output, the efficiency wedge seems to matter the most and the differences are statistically significant. The efficiency wedge economy is also the only for which we can reject the null hypothesis that the success ratio (equal to 68%) is smaller or equal to 50%. As in the case of the RMSEs, this is also confirmed for the all-but-one-wedge economies.

4.2 Implications for modeling financial frictions

The analysis of simulations for economies with just one wedge and economies with all but one wedge provide robustness to previous findings that stress the importance of modeling distortions that resemble TFP shocks in order to replicate movements in output and investment, and labor income taxes, in order to replicate movements in hours. Distortions to the savings decision,

i.e., extensions that can be mapped to the investment wedge, seem of little promise both quantitatively and qualitatively.

Do these results suggest that financial frictions are irrelevant? One should have in mind that the financial system has two functions. One is to allow agents to transfer resources across time and states of the world. The other is to channel savings to their most efficient use. Classical model of financial frictions, such as Bernanke et al. (1999) and Calstrom and Fuerst (1997), typically produce distortions to the equilibrium conditions of the neoclassical growth model that express themselves as investment wedges. What these results show is, in line with what Chari et al. (2007) showed for the U.S. economy for these two examples, this generation of models holds little promise in generating fluctuations in observables as in the data. In a representative agent framework, distortions to the ability of transferring resources across time or states of the world by themselves, as in the costly state verification mechanism in the two examples cited above, pale in comparison with the quantitative and qualitative relevance of models of the labor wedge or total factor productivity. In a representative agent framework, the most promising theories of financial frictions will describe mechanisms that induce inefficiencies in resource allocation or labor markets.

In work by Kehoe and McGrattan (2005) we can find examples on how financial frictions in the form of advance payment constraints can appear as labor and efficiency wedges. The authors show that in an environment where firms need to set aside the one period ahead wage bill in a non-interest bearing account, will reflect that cost in the wage and therefore create a wedge between the marginal productivity of labor and the wage. In the same paper the authors also show how an advance payment-payment constraint on intermediate goods coupled with collateral constraints will express itself as total factor productivity shocks. In another example, recent work by Arellano et al. (2012) focuses on a model with advance payment constraints, uninsurable idiosyncratic shocks and increased volatility at the firm level to explain the decline in output during the Great Recession. As in Kehoe and McGrattan (2005), these frictions also express themselves as labor wedges. Work by Buera and Moll (2012) show that credit constraints in an

heterogeneous agents environment induce distortions that once aggregated can express themselves as efficiency or labor wedges in the representative agent framework. Chang and Kim (2007) show, also in an environment with heterogeneous agents, that incomplete markets and indivisible labor aggregate also to distortions to the intra-temporal condition. More examples can be found in the works of Perri and Quadrini (2011) and Lopez (2011).

As a final note it is worth noting that the model's ability to replicate output and hours is much higher compared with investment. As seen in Figures 1-6 in Appendix B, the average absolute simulation errors for deviations from the trend are one order of magnitude larger. This is also confirmed by the RMSEs. As an example, the RMSEs for output and hours in the efficiency wedge economy are of 1.33 and 1.66, against 9.74 for investment. A relationship of the same magnitude can be observed for the other types of economies (see Tables 1-2 and Figures 7-12 in Appendix B).

One of the reasons why investment dynamics are harder to replicate may have to do with mismeasurement in the data. McGrattan and Prescott (2010) shows how unmeasured investment (and output) may have biased output and productivity estimates in the 1990s downwards. By introducing a two sector model with non-neutral technological and intangible investment, the authors manage to correct counterfactual predictions standard real business cycle theory produces in that period.

5 Structural analysis of the wedges

The ultimate goal of a business cycle accounting exercise is to enable practitioners to come forth with structural explanations for the wedges. If, for example, the labor wedge is quantitatively relevant for a period of fluctuations, then researchers should focus in developing extensions to the prototype economy that, if mappable to the labor wedge, can replicate it. The current section contributes to the literature by bringing forth factors that contain information with respect to the said wedges.

5.1 Spatial correlation

Table 5 shows that the wedges' correlations of the cyclical components across countries is correlated with the geographical distance between them (as measured by the geographical distance between each country's capital), in line with what Aguiar-Conraria and Soares (2011) find concerning business cycle synchronization in the Euro-Area. In other words, the closer two countries are, the more correlated their efficiency wedges will be, for example.

Table 5: Distance vs Correlation between wedges

	A_t	$\tau_{l,t}$	$\tau_{x,t}$	g_t
AUS	-0.42*	-0.36	-0.14	-0.14
BEL	-0.56**	-0.38	-0.45*	-0.38
CAN	-0.40*	-0.14	-0.14	-0.02
CHE	-0.46**	-0.24	-0.33	-0.27
DNK	-0.49**	-0.22	-0.49**	-0.34
ESP	-0.51**	0.02	-0.18	-0.06
FIN	-0.51**	-0.10	-0.37	-0.07
FRA	-0.67***	-0.25	-0.57**	-0.34
GBR	-0.38	0.11	-0.49**	-0.02
ISL	-0.44*	0.03	-0.13	-0.08
ITA	-0.79***	-0.17	-0.47**	-0.24
JPN	0.08	0.01	-0.38	-0.17
KOR	0.03	-0.24	0.10	0.00
LUX	-0.77***	-0.40*	-0.34	-0.31
NLD	-0.57**	-0.60***	-0.33	-0.48**
NOR	-0.49**	-0.14	-0.10	-0.26
NZL	0.30	-0.38	0.14	-0.20
SWE	-0.75***	-0.36	-0.48**	-0.41*
USA	-0.33	-0.22	0.15	-0.28

Statistical significance levels for Spearman's rank order correlation coefficient

*** < 0.01, ** < 0.05, * < 0.10

The values correspond to Spearman's rank correlation coefficient for each of the wedges across countries (and its statistical significance). In only 11 cases out of 76 we see a positive association between the distance across two countries and the correlation coefficient between their corresponding wedges. However for not a single one of these 11 cases, we observe statistical significance.

It is thus suggested that the further two countries are apart, the lesser the wedges are correlated. This effect is especially strong in the case of the efficiency wedge, where many correlations found are indeed statistically significant.

This is in line with the idea behind gravity type of instruments. If trade is one channel through which we can observe the international transmission of shocks, and if distance increases trade costs and therefore make countries trade less, then one would expect the correlations to be negative, as indeed it is for most cases in Table 5. Cavallo and Frankel (2008) make use of this mechanism to study the impact of the degree of openness to trade in the likelihood of countries experiencing sudden stops or currency crashes.

5.2 Degree of Openness to Trade

The following table brings forth a second factor with potential to explain fluctuations in the wedges at the business cycle frequency. A country's openness to trade is defined as the (HP-filtered) size of the sum of real exports and real imports as a share of real GDP.

In Table 6 we provide evidence on the information that our measure of openness contains with regard to the measured wedges. With regard to TFP or the efficiency wedge, in 15 out of 19 countries, the correlations are significant at least at the 5% level. In all of those 15 cases, the correlation is positive. The effect of openness in growth has been perceived in the literature as positive (Frankel and Romer (1999)). Output growth, however, can under the neoclassical model, be attributed to an increase in either capital, labor or total factor productivity. Much work has focused in the effects of trade openness on total factor productivity growth and Abizadeh and Pandey (2009) find these effects to be positive, which is consistent with the results in Table 6. Another channel through which the increase in foreign direct investment can also be positively associated with gains in efficiency is by the pressure that may be put upon local officials to remove domestic distortions.

Table 6: Trade openness correlation with the wedges

	A_t	$\tau_{l,t}$	$\tau_{x,t}$	g_t
AUS	0.02	-0.41***	0.10	-0.43***
BEL	0.61***	-0.41***	0.67***	-0.46***
CAN	0.63***	-0.56***	0.30***	-0.46***
CHE	0.49***	0.02	0.22***	0.00
DNK	-0.06	-0.15*	0.14*	-0.02
ESP	0.36***	-0.47***	0.02	-0.57***
FIN	0.52***	-0.07	0.29***	0.05
FRA	0.61***	-0.44***	0.50***	-0.60***
GBR	0.20***	-0.30***	0.20***	-0.27***
ISL	0.18**	-0.25***	-0.14*	-0.42***
ITA	0.62***	-0.04	0.23***	-0.12
JPN	0.38***	-0.21***	0.14*	0.01
KOR	0.37***	-0.26***	-0.28***	-0.27***
LUX	0.32***	0.23***	0.07	0.19**
NLD	0.38***	-0.07	0.20***	-0.09
NOR	0.03	-0.17**	0.11	-0.21***
NZL	-0.10	-0.33***	0.04	-0.36***
SWE	0.45***	-0.24***	0.19**	-0.22***
USA	0.44***	-0.67***	-0.51***	-0.50***

Statistical significance levels for Spearman's rank order correlation coefficient

*** < 0.01, ** < 0.05, * < 0.10

In the case of the government wedge, 13 correlations are found to be significant (at least at the 5% level) and all but one - again, Luxembourg - to be negative. In a paper by Cavallo and Frankel (2008) mentioned before, trade openness leads to a lesser probability that sudden stops or currency crisis may occur. As Kehoe and McGrattan (2005) show, sudden stops, i.e., an abrupt decrease in new borrowing, will appear in the prototype economy, as a surge in exports or an increase in the government wedge. In light of this mechanism, one would indeed expect a negative correlation between trade openness and the government wedge as observed in Table 6.

Modeling choices of sudden stops have also suggested mechanisms that show as labor and efficiency wedges, thus leading to a rationality as why trade variables and the wedges would be related. Neumeyer and Perri (2005) and Mendoza (2010) are but a couple of examples. With regard to the labor wedge, all significant (at least at the 10% level) correlations are negative,

except one - Luxembourg. As in the case before, this happens for 15 out of the 19 countries in the sample. For the investment wedge, fewer correlations (14) are found to be statistically significant and 11 to be positive.

There is a large increase in trade and foreign direct investment during the 1980's. To assess the robustness of the results, I compute the same correlations for two periods, from 1970 to 1985 and from 1986 onwards. The results are fairly robust in terms of number of significant correlations (57 vs 47 for the 1970 to 1985 period and 57 vs 55). In terms of qualitative changes, there were only 5 cases where significant correlations changed to significant correlations of the opposite sign. Four of those were with regard to Denmark and Japan. In terms of magnitude of the correlations, they are about the same for the investment wedge, but there is a significant increase for the other wedges' correlation with our measure of openness to trade. In the case of the efficiency wedge for example, the increase in the average correlation is from 0.24 to 0.41.

6 Conclusion

This paper studies the properties of distortions to the neoclassical growth model that manifest themselves as shocks productivity shocks, labor income taxes, investment taxes, and government consumption. It is shown that deviations from the trend with regard to these distortions are relatively more synchronized in the case of the efficiency, labor and government wedge than in the case of the investment wedge. Recessions in the U.S. typically coincide with local extreme points in the wedges for the aggregate data and these movements are similar enough to be statistically significant in the case of the efficiency, labor and government wedges. This is even more remarkable given that the data used is unweighted and shows the impact the U.S. hold for business cycle fluctuations for the rest of the OECD countries in the sample.

The relevance of the labor wedge relative to total factor productivity in the U.S. during the Great Recession is also a distinctive feature of the U.S. when compared to most Western European countries. While in the U.S. the recession is associated with a much larger labor wedge drop, in

Western Europe most countries experienced a greater decrease in TFP, a result consistent with Ohanian and Raffo (2012).

Simulations show that the efficiency wedge is paramount in providing models the ability to replicate movements in output and investment. With regard to output, an economy with just an efficiency wedge outperforms, on average, an economy with all the other three wedges. The labor wedge is also of relevance in replicating movements in hours and investment where the investment wedge holds little to no qualitative or quantitative importance. These conclusions are statistically significant for the countries in the sample and do not limit themselves to periods such as output drops, as it has been traditionally done in the literature.

The original findings of Chari et al. (2007) with respect to the irrelevance of the investment wedge are therefore reinforced by these results and are also valid for business cycle fluctuations in general. However, how to reconcile the general notion that many of the recent recessions were due to failures of the financial system with such irrelevance? One should draw attention to the fact that the financial system has two purposes. First it allows the allocation of resources across time and states of the world. It allows consumers to smooth their consumption in a context of variable income, or firms to stabilize their liquidity in face of revenue/cost shocks. The second function is to channel savings to their most productive uses. The first purpose is reflected in the model by the Euler equation and its associated wedge is precisely the investment wedge. The second function of the financial markets tends to be captured by the efficiency wedge. As mentioned before, Chari et al. (2007) describe a detailed two sectors economy with different costs of borrowing. In this case, they derive an equivalence theorem between the detailed economy and a prototype economy with an efficiency wedge.

The conclusion here is that promising theories of financial crisis should focus on the role of the financial system in channeling savings to their most efficient uses, and less so in its role of allocating resources across time, at least in the representative agent framework. The issue of liquidity crisis at the aggregate level is not persistent enough across time, and especially hard to capture at the quarterly frequency even in studies of financial crisis which

has been typically the aim of much of the literature. Even in models with advance payment constraints, as shown by Kehoe and McGrattan (2005), it is not the timing of the payment that generates a distortion, but the need that funds are held in an escrow account. An exception is precisely Dooyeon and Doblus-Madrid (2012) that find some relevance of the investment wedge for a selected sample of east Asian economies, but again, limited to financial crisis. In a context of distortions that aim to generate fluctuations in aggregates that are not limited to output drops or crisis, the relevance of such channel is expected to be even less important.

Also, I show that the per type of wedge across country correlation is negatively correlated with the distance between the countries, even though only for the efficiency wedge, strong statistical significance can be found. As mentioned before, this type of analysis is common in the trade literature, where trade volumes are often (also) explained by gravitation equations. The degree of openness to trade of a country is also found to contain significant information with regard to all four wedges which reinforces the role of international mechanisms of transmission for distortions in the key margins of the neoclassical growth model.

Further analysis should focus in finding more factors that contain information with regard to these distortions, so that we can provide business cycle theorists hints of mechanisms that hold promise in generating fluctuations as observed in the data.

References

- Abizadeh, S. and Pandey, M. (2009). Trade openness, structural change and total factor productivity. *International Economic Journal*, 23(4):545–559.
- Aguiar-Conraria, L. and Soares, M. J. (2011). Business cycle synchronization and the Euro: A wavelet analysis. *Journal of Macroeconomics*, 33:477–489.
- Arellano, C., Bai, Y., and Kehoe, P. J. (2012). Financial frictions and fluctuations in volatility. *Federal Reserve Bank of Minneapolis Research Department Staff Report*, 466.
- Bernanke, B. M., Gertler, M., and Gilchrist, S. (1999). The financial accelerator in a quantitative business cycle framework. *Handbook of Macroeconomics*, 1C:1341–1393.
- Brinca, P. (2013). Monetary business cycle accounting for sweden. *The B.E. Journal of Macroeconomics*, 13(1):2194–6116.
- Buera, F. J. and Moll, B. (2012). Aggregate implications of a credit crunch. Working Paper 17775, National Bureau of Economic Research.
- Calstrom, C. T. and Fuerst, T. S. (1997). Agency costs, net worth and business cycle fluctuations: a computable general equilibrium analysis. *American Economic Review*, 87:893–910.
- Cavallo, E. A. and Frankel, J. A. (2008). Does openness to trade make countries more vulnerable to sudden stops, or less? using gravity to establish causality. *Journal of International Money and Finance*, 27(8):1430–1452.
- Chang, Y. and Kim, S.-B. (2007). Heterogeneity and aggregation: Implications for labor-market fluctuations. *The American Economic Review*, 97(5):1939–1956.
- Chari, V., Kehoe, P., and McGrattan, E. (2007). Business cycle accounting. *Econometrica*, 75:781–836.

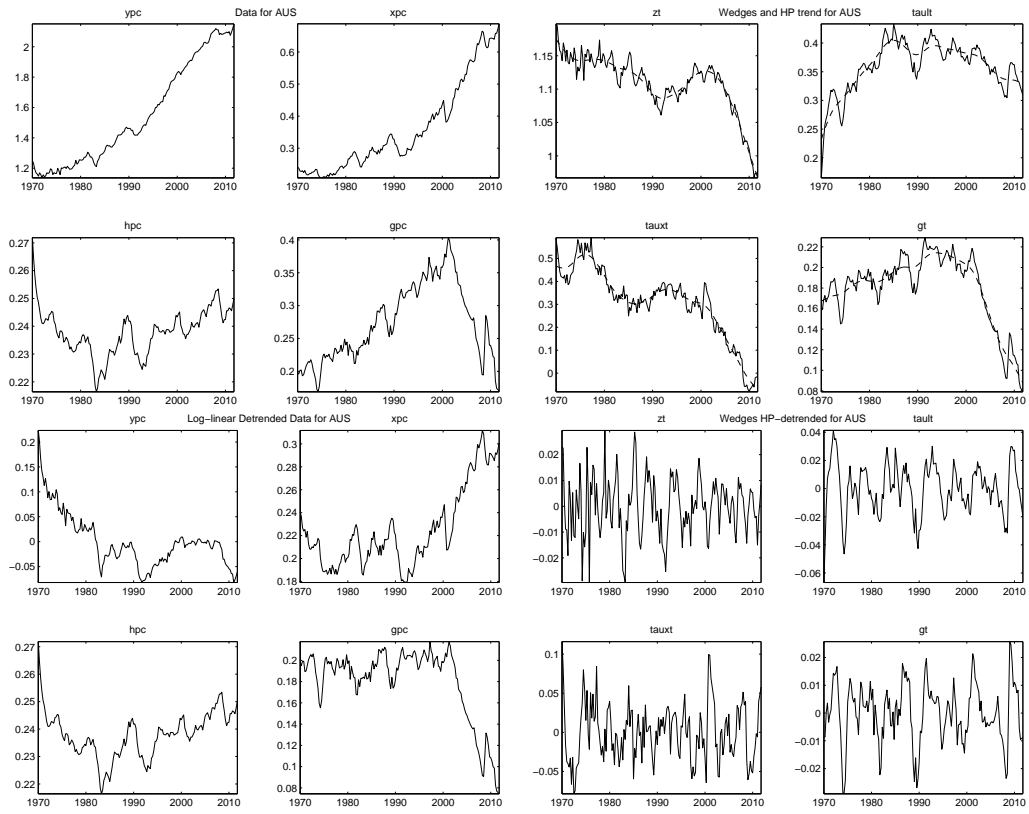
- Chari, V., Kehoe, P. J., and McGrattan, E. R. (2005). Sudden stops and output drops. *American Economic Review*, 95(2):381–387.
- Cociuba, S. E. and Ueberfedt, A. A. (2010). Trends in U.S. hours and the labor wedge. *Bank of Canada Working Paper Series*, (28).
- Dooyeon, C. and Doblas-Madrid, A. (2012). Business cycle accounting east and west: Asian finance and the investment wedge. *Review of Economic Dynamics*. Available online 17 October 2012, ISSN 1094-2025, 10.1016/j.red.2012.10.003.
- Frankel, J. A. and Romer, D. (1999). Does trade cause growth? *American economic review*, pages 379–399.
- Hoddrick, R. and Prescott, E. (1997). Post-war U.S. business cycles: an empirical investigation. *Journal of Money, Credit and Banking*, 29:1–16.
- Islam, N., Dai, E., and Sakamoto, H. (2006). Role of TFP in China’s growth. *Asian Economic Journal*, 20(2):127–156.
- Kehoe, T. J. and Prescott, E. C. (2007). *Great depressions of the twentieth century*. Federal Reserve Bank of Minneapolis.
- Kobayashi, K. and Inaba, M. (2006). Business cycle accounting for the Japanese economy. *RETI Discussion Paper*, 05-E-023.
- Kose, M. A., Otrok, C., and Prasad, E. (2012). Global business cycles: Convergence or decoupling? *International Economic Review*, 53(2):511–538.
- Kydland, F. E. and Prescott, E. C. (1982). Time to build and aggregate fluctuations. *Econometrica*, 50(6):1345–70.
- Lamas, R. (2009). Accounting for output drops in Latin America. *IMF Working Paper Series*, 67.
- Lopez, J. I. (2011). Consumption and labor income risk, aggregation and business cycles.

- McGrattan, E. R. and Prescott, E. C. (2009). *Technical appendix: Unmeasured investment and the puzzling US boom in the 1990s*.
- McGrattan, E. R. and Prescott, E. C. (2010). Unmeasured investment and the puzzling us boom in the 1990s. *American Economic Journal: Macroeconomics*, 2(4):88–123.
- Mendoza, E. G. (2010). Sudden stops, financial crises, and leverage. *The American Economic Review*, 100(5):1941–1966.
- Mertens, K. and Ravn, M. (2008). The aggregate effects of anticipated and unanticipated us tax policy shocks: Theory and empirical evidence. *CEPR Discussion Paper No. DP6673*.
- Mulligan, C. B. (2002). A century of labor-leisure distortions. *NBER Working Papers*, (8774).
- Neumeyer, P. A. and Perri, F. (2005). Business cycles in emerging economies: the role of interest rates. *Journal of monetary Economics*, 52(2):345–380.
- Ohanian, L., Raffo, A., and Rogerson, R. (2008). Long-term changes in labor supply and taxes: Evidence from OECD countries, 1956–2004. *Journal of Monetary Economics*, 55(8):1353–1362.
- Ohanian, L. E. and Raffo, A. (2012). Aggregate hours worked in oecd countries: New measurement and implications for business cycles. *Journal of Monetary Economics*, 59(1):40–56.
- Otsu, K. (2009). International business cycle accounting. *IMES Discussion Paper Series*, 29.
- Perri, F. and Quadrini, V. (2011). International recessions. Technical report, National Bureau of Economic Research.
- Restrepo-Echavarria, P. and Cheremukhin, A. A. (2010). The labor wedge as a matching friction. *Federal Reserve Bank of Dallas Working Paper Series*, (1004).

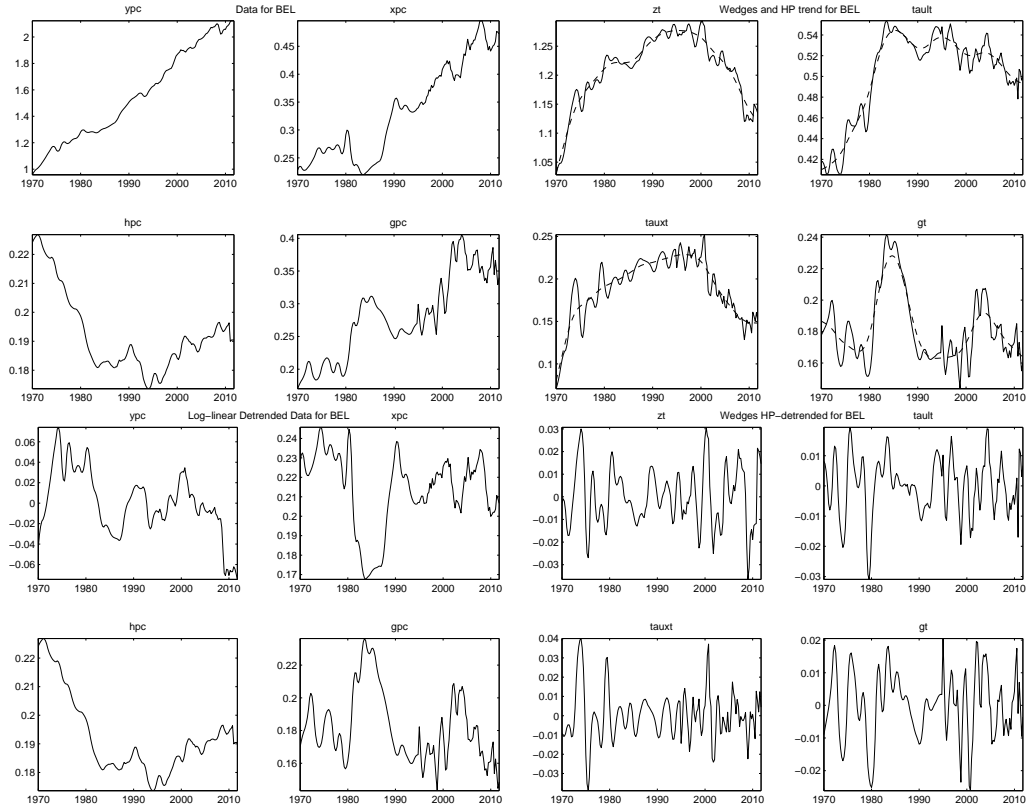
- Samuelson, P. A. (1954). The transfer problem and transport costs, ii: Analysis of effects of trade impediments. *The Economic Journal*, 64(254):264–289.
- Shimer, R. (2009). Convergence in macroeconomics: The labor wedge. *American Economic Journal: Macroeconomics*, 1(1):280–97.
- Shimer, R. (2010). *Labor Markets and Business Cycles*. Princeton University Press.
- Simonovska, I. and Söderling, L. (2008). Business cycle accounting for Chile. *IMF Working Paper Series*, 61.
- Solow, R. M. (1957). Technical change and the aggregate production function. *The Review of Economics and Statistics*, 39(3):312–320.
- Sustek, R. (2010). Monetary business cycle accounting. *Review of Economic Dynamics*, 14(4):592–612.

Appendix A - Country Profiles

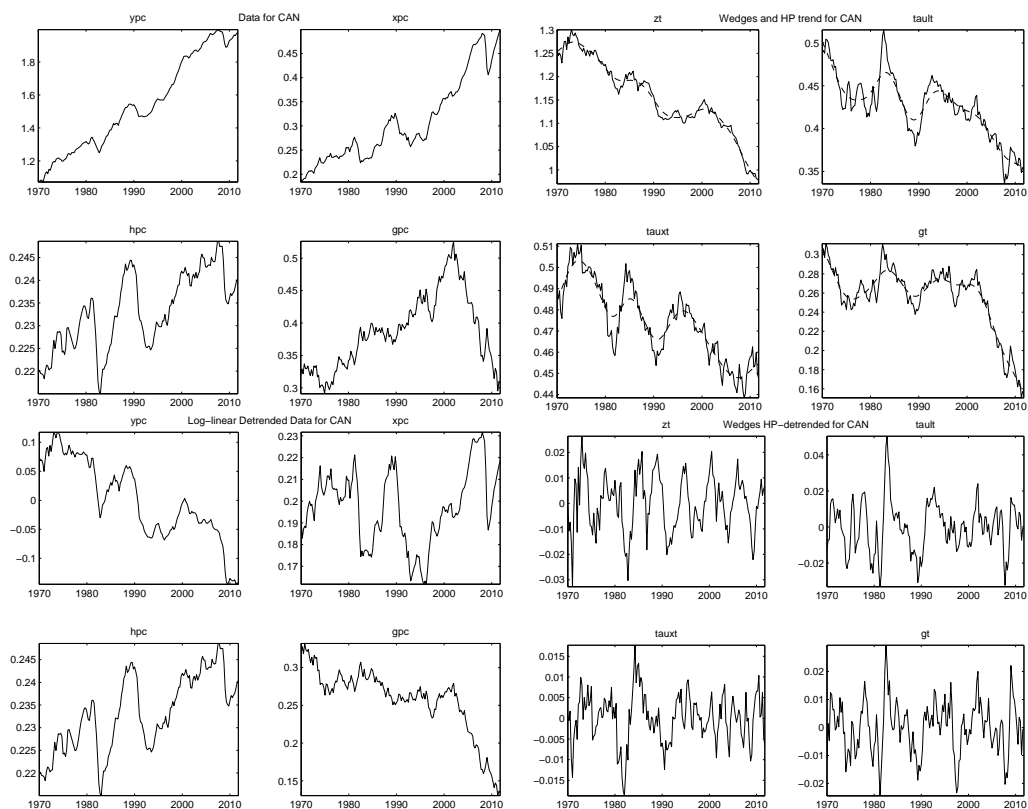
Australia



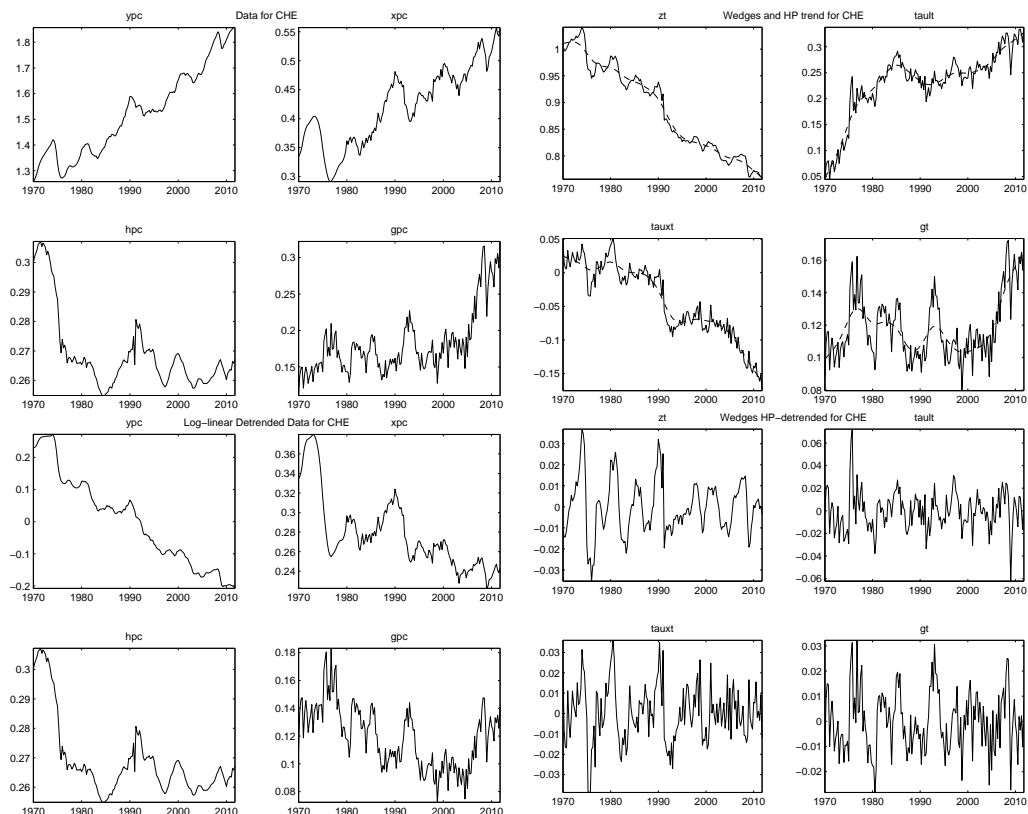
Belgium



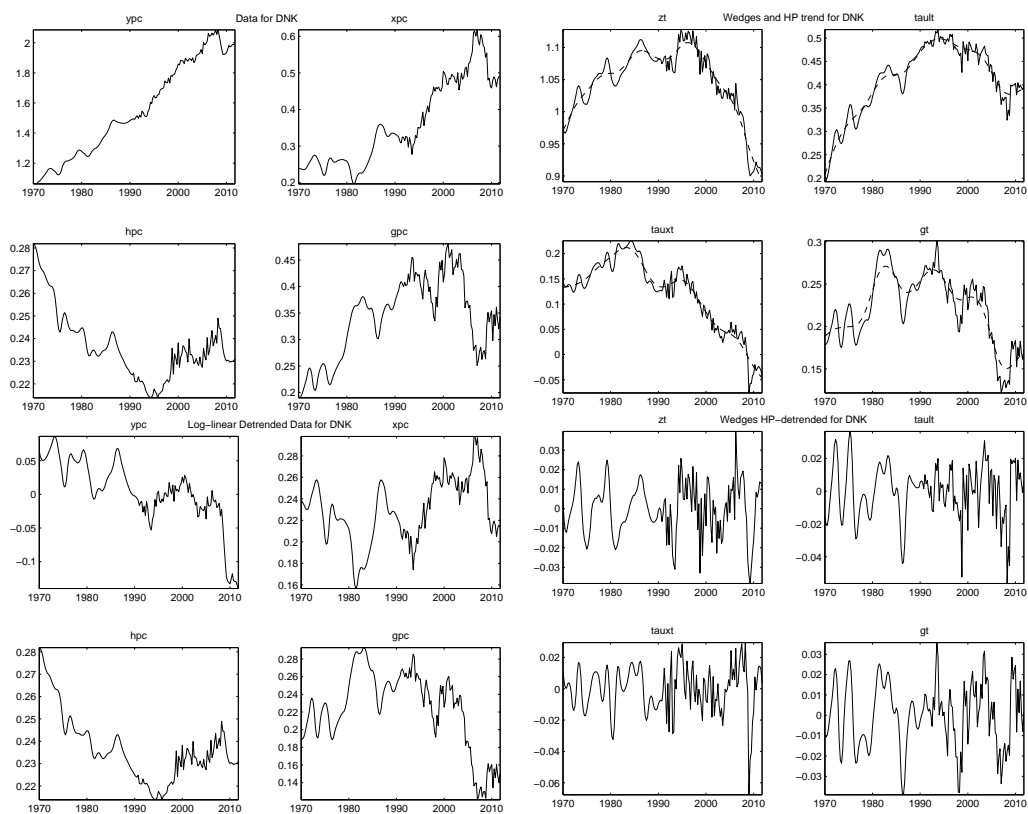
Canada



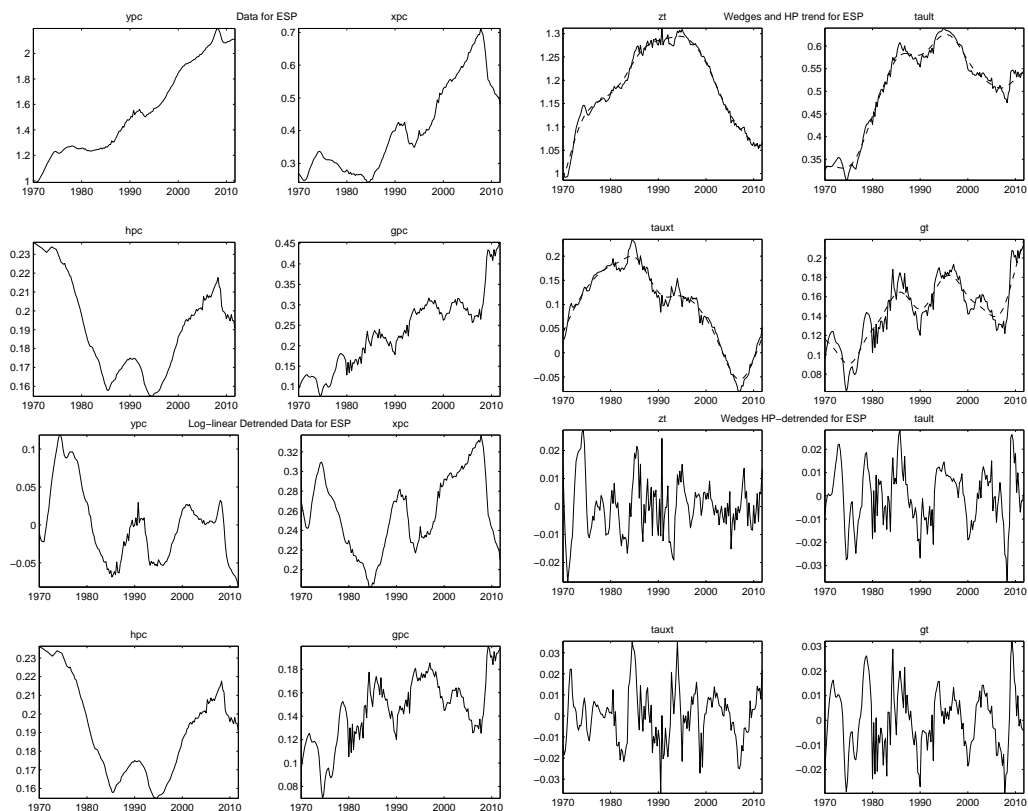
Switzerland



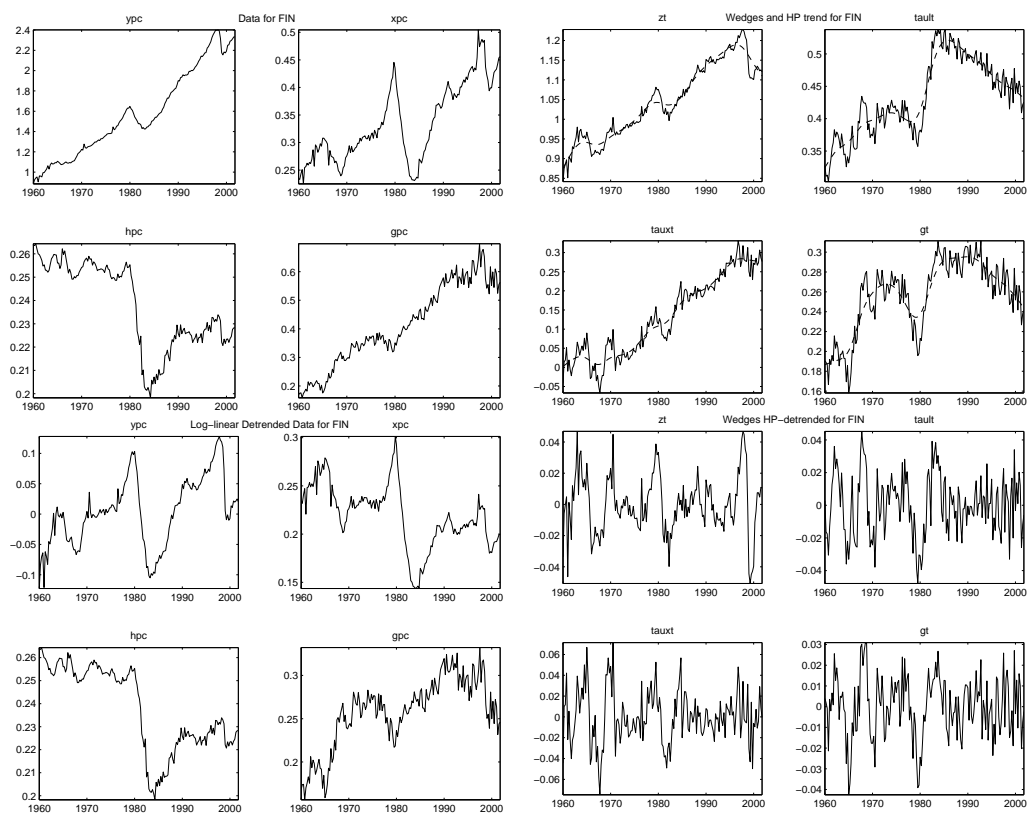
Denmark



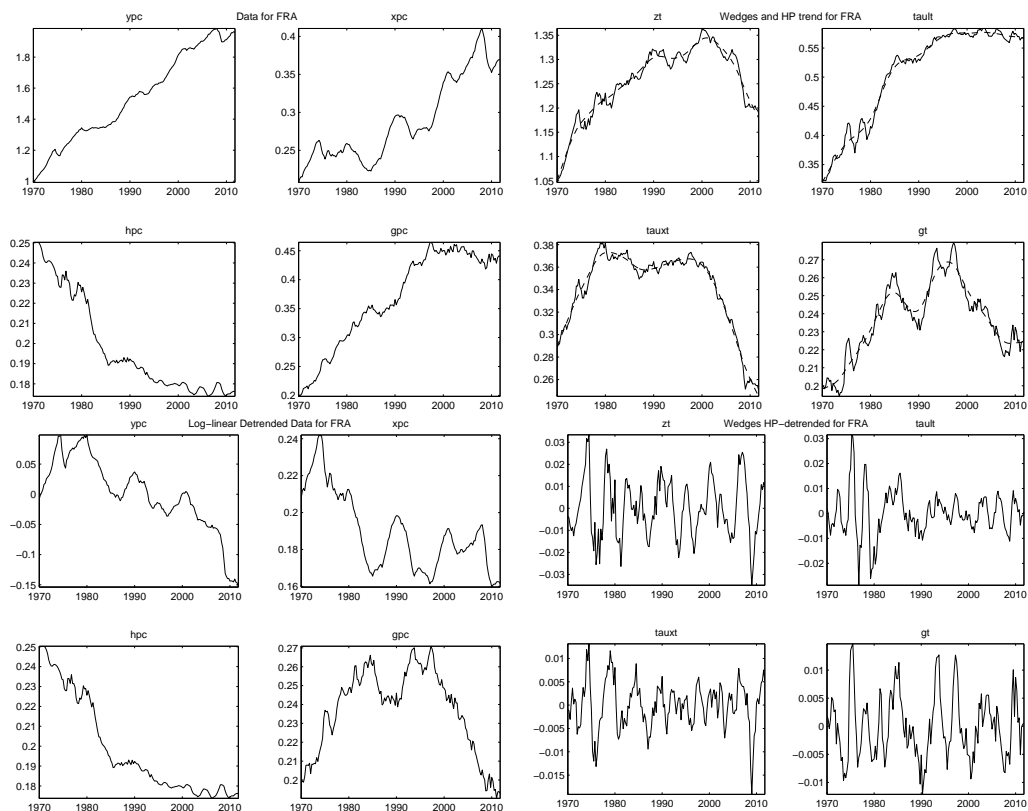
Spain



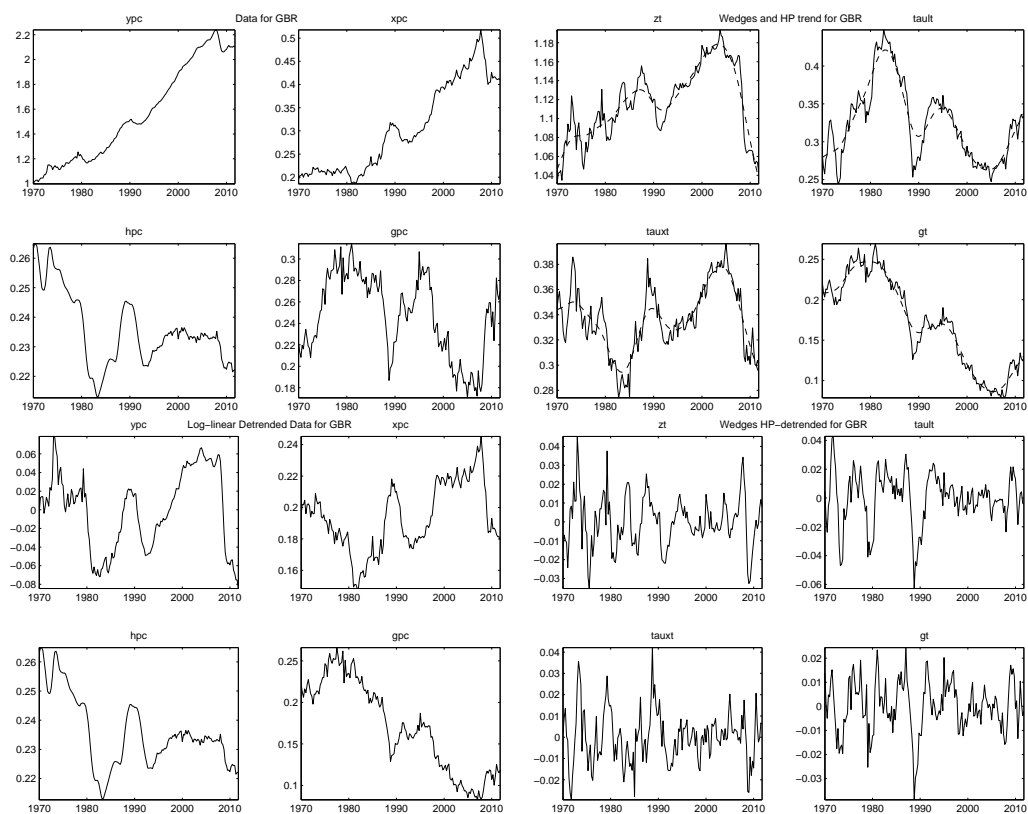
Finland



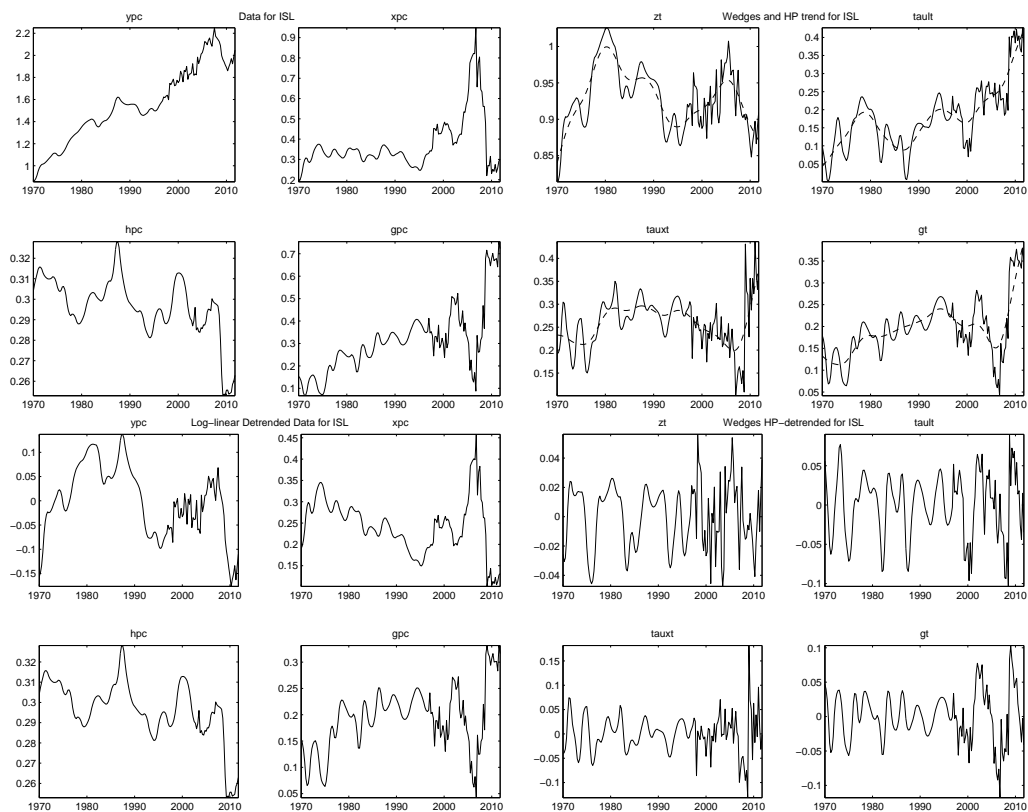
France



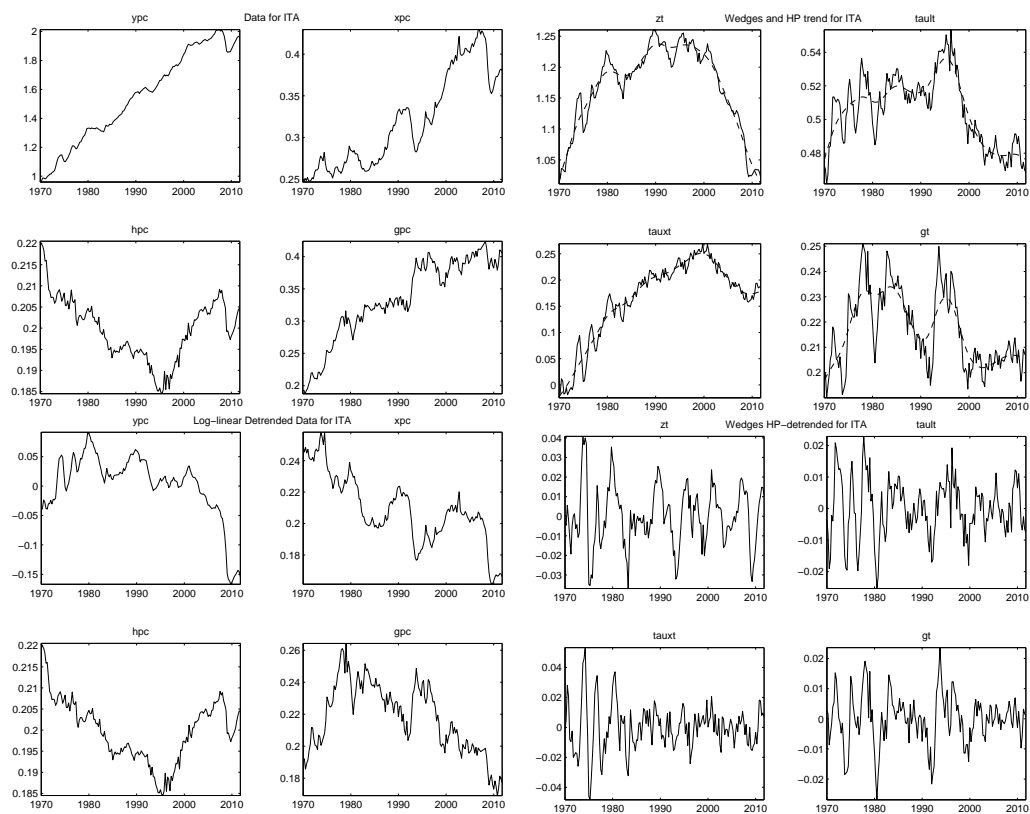
Great Britain



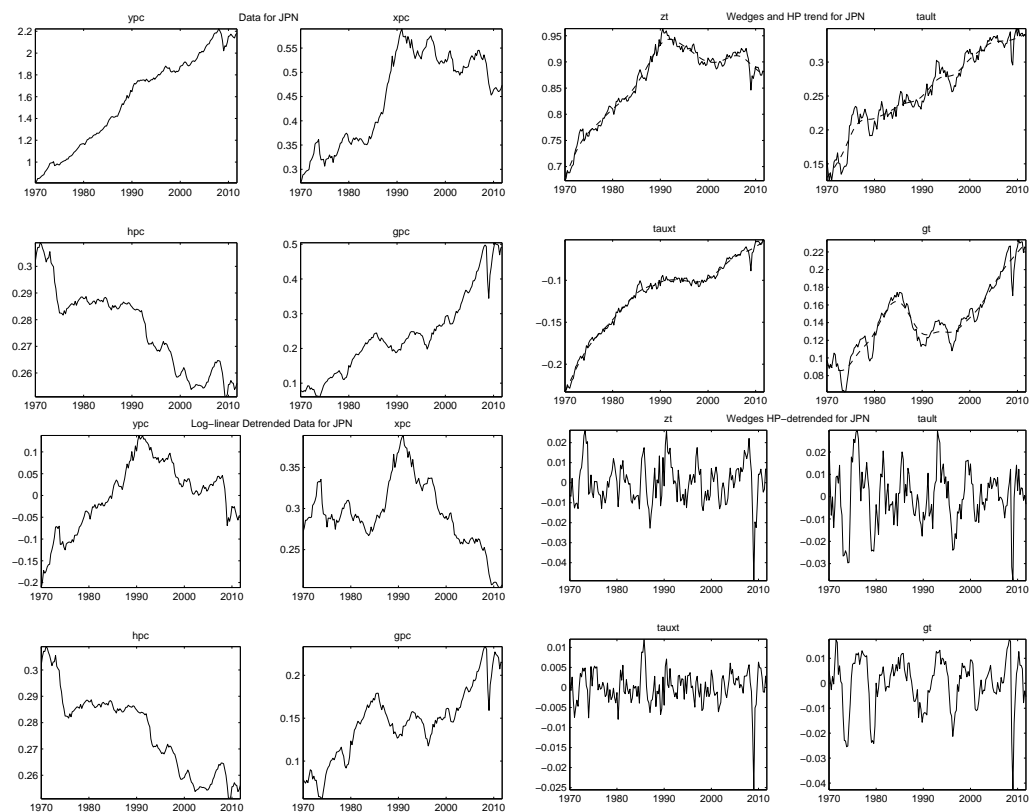
Iceland



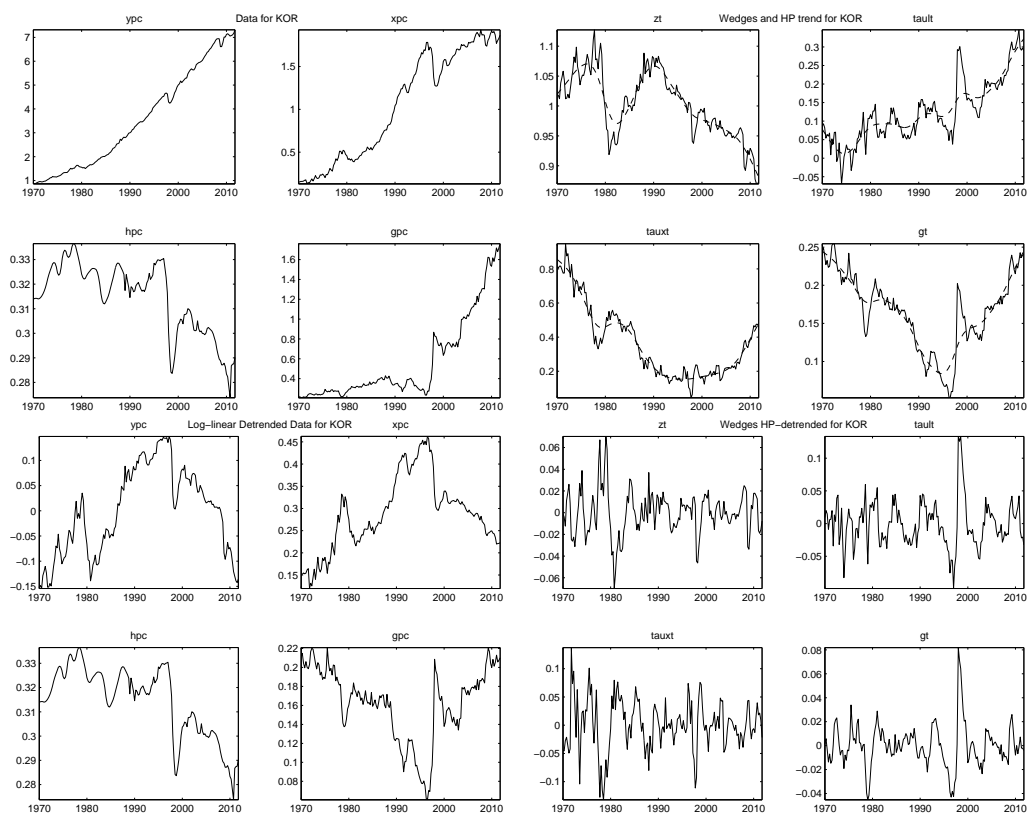
Italy



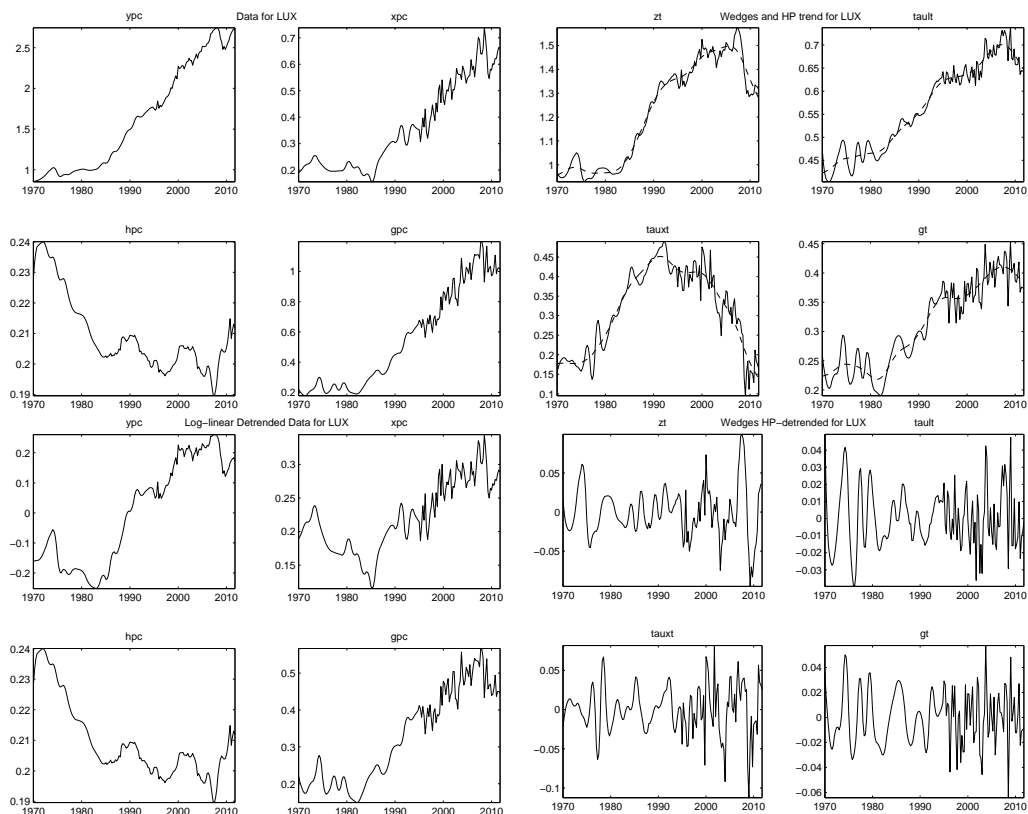
Japan



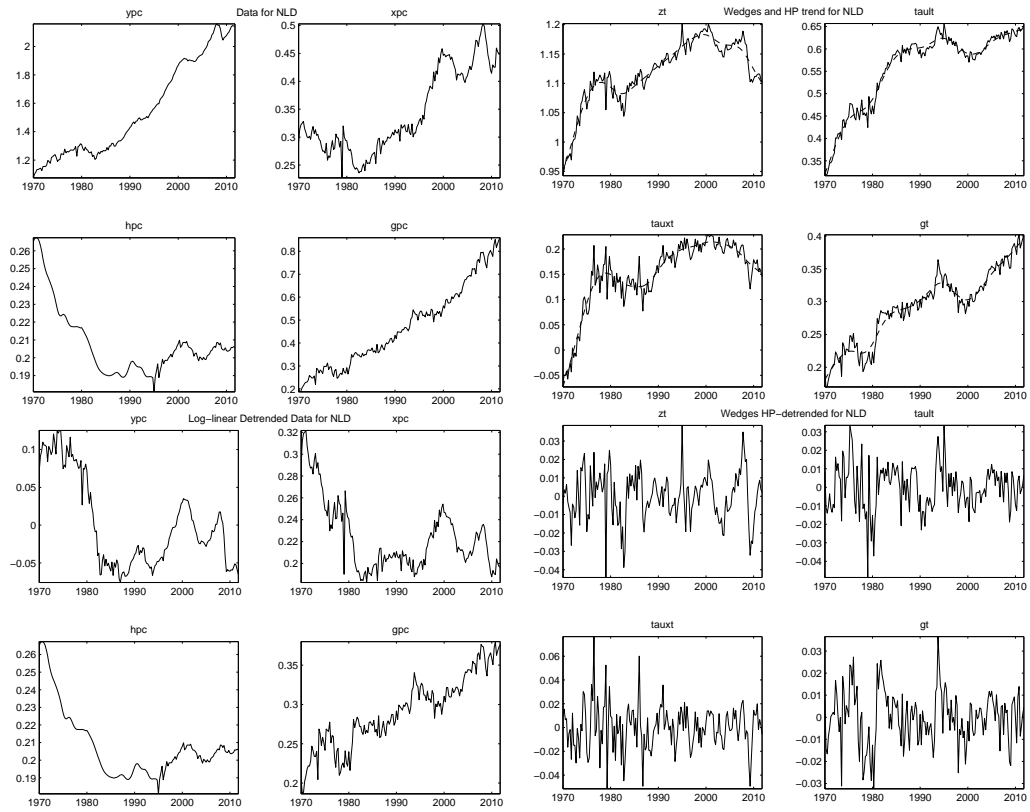
South Korea



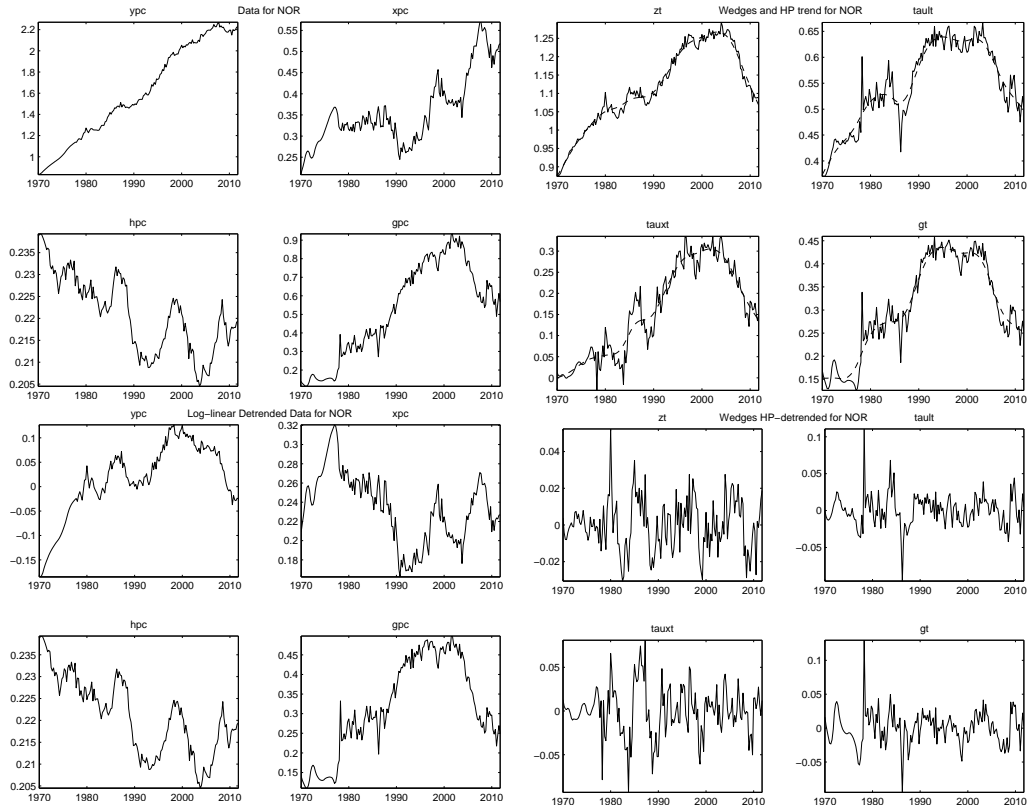
Luxembourg



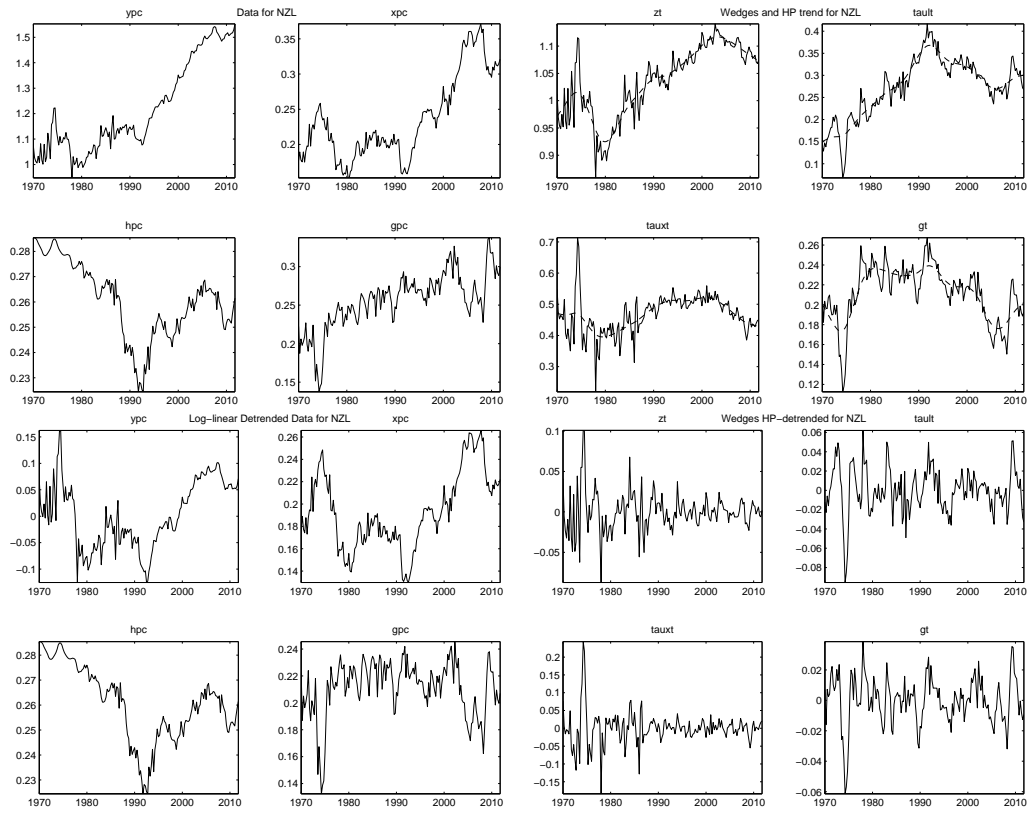
Netherlands



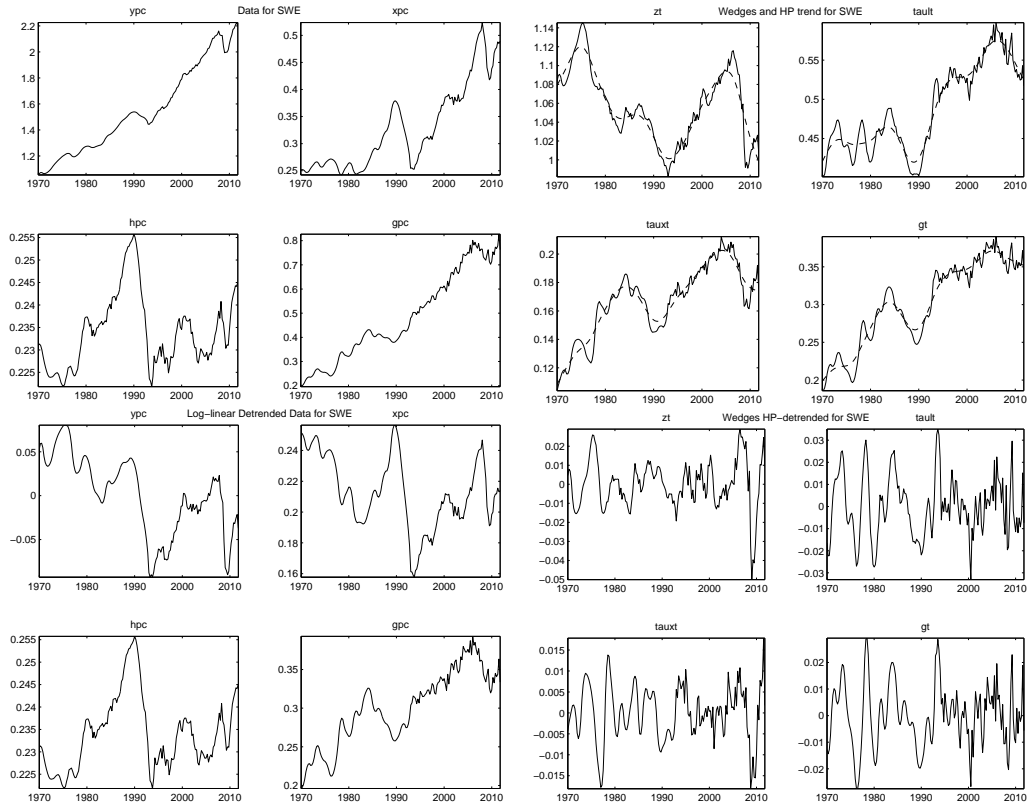
Norway



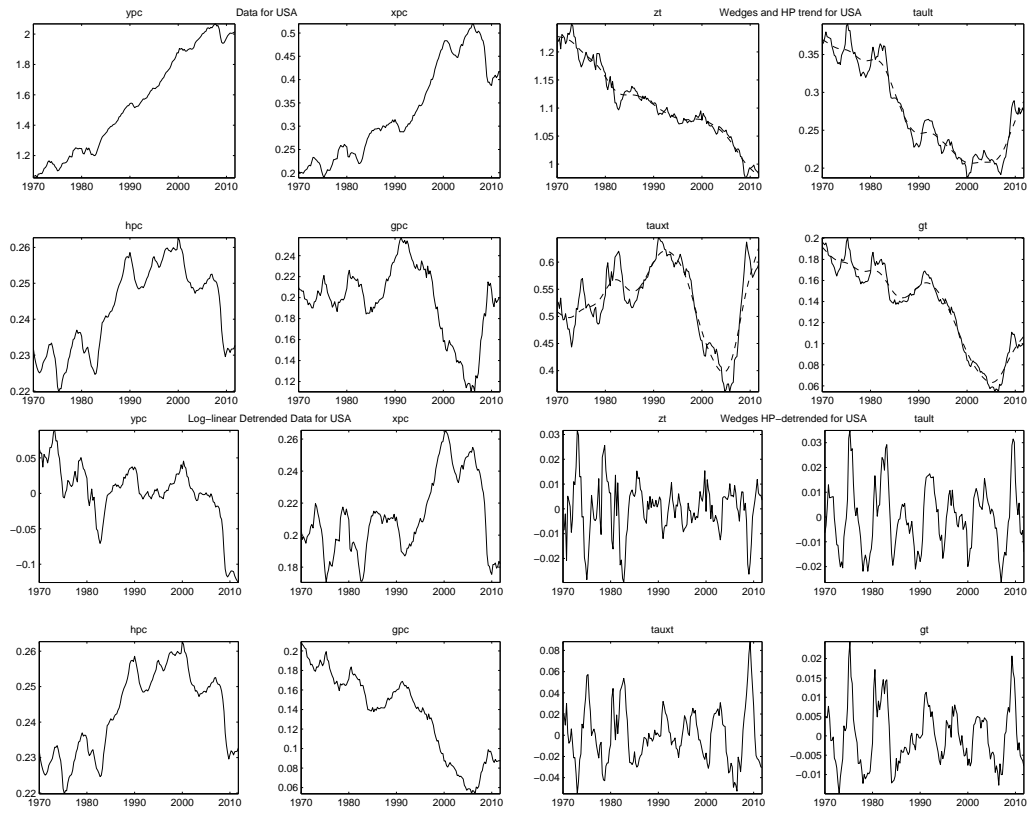
New Zealand



Sweden



USA



Appendix B - Simulations

Deviations from Cycle for 1 Wedge and 3 Wedge Economies

Figure 1: Output

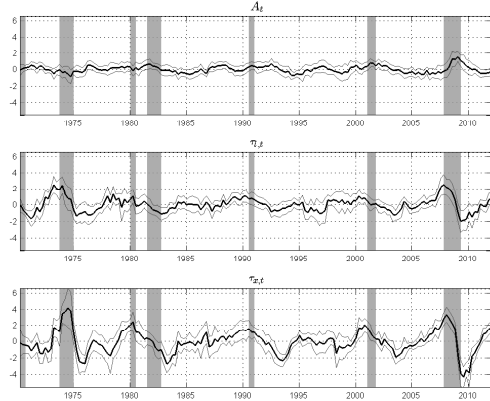


Figure 4: Output

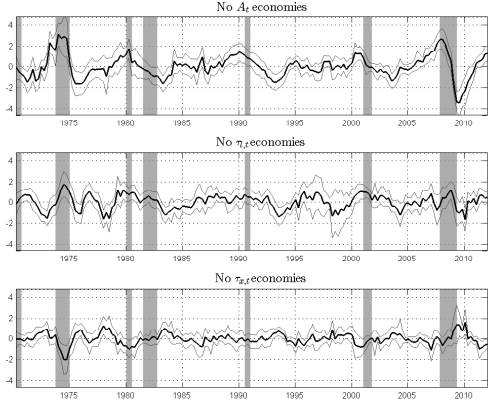


Figure 2: Hours worked

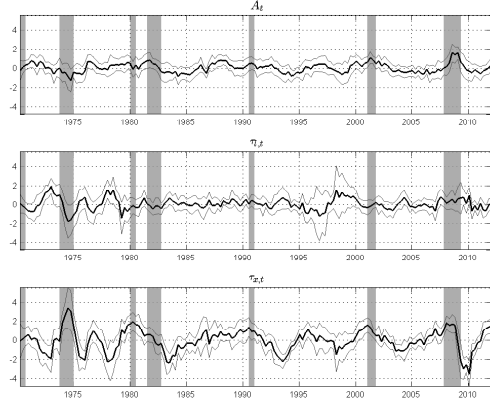


Figure 5: Hours worked

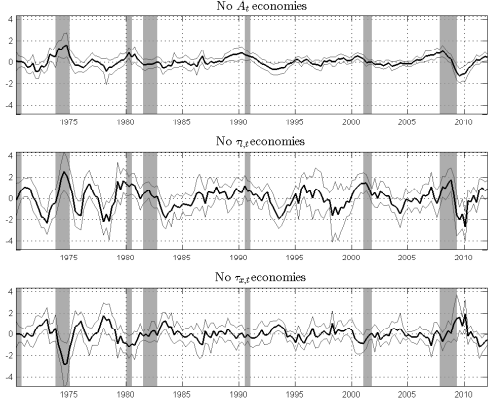


Figure 3: Investment

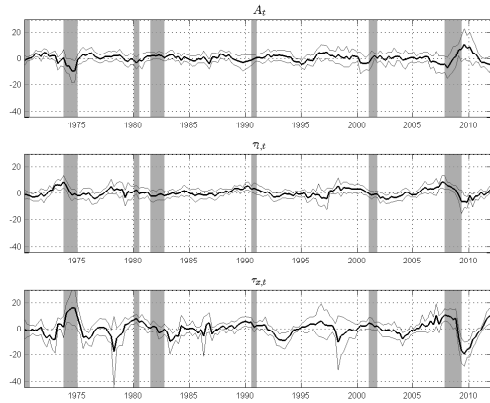
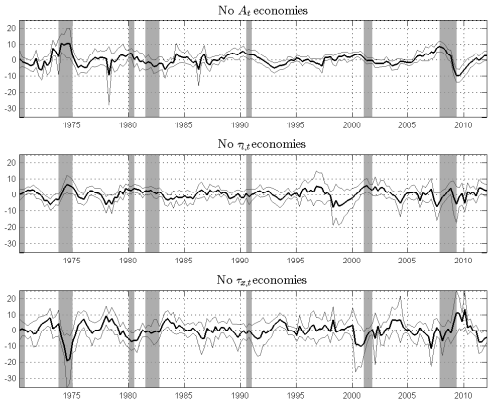


Figure 6: Investment



RMSEs for 1 Wedge and 3 Wedge Economies

Figure 7: Output

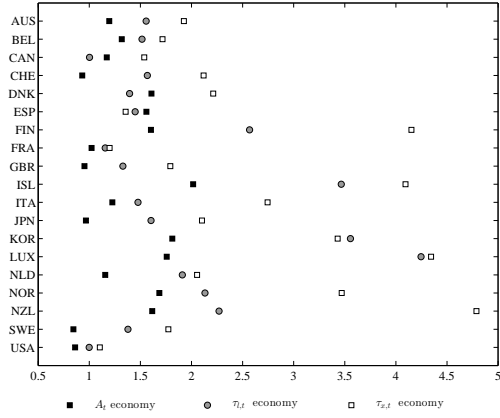


Figure 10: Output

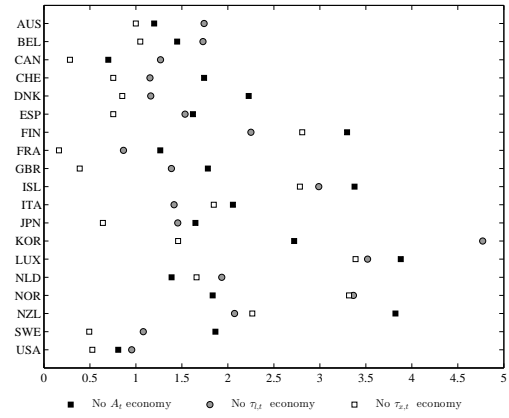


Figure 8: Hours worked

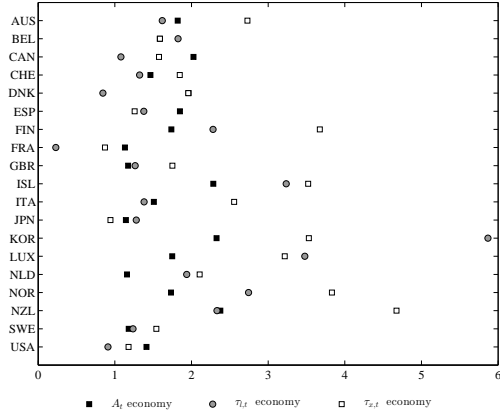


Figure 11: Hours worked

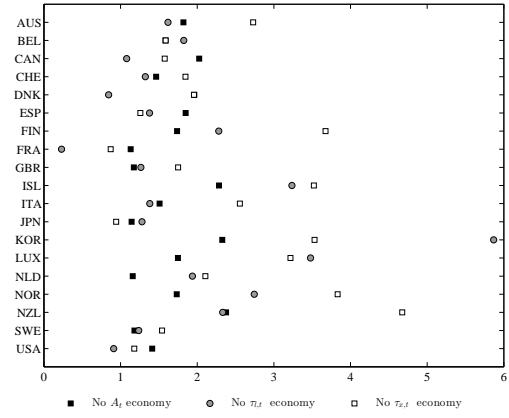


Figure 9: Investment

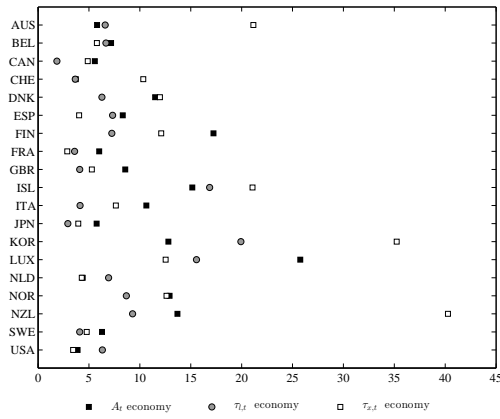
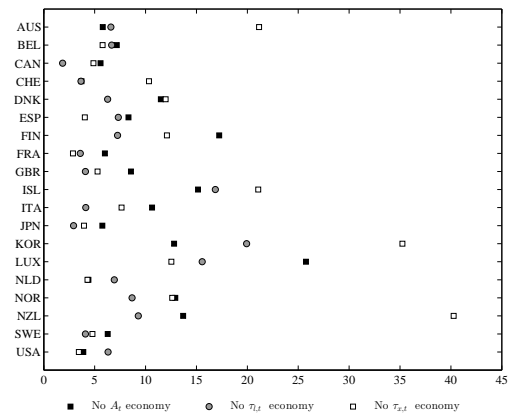


Figure 12: Investment



Statistical Tests of Comparative Performance - RMSEs

Table 1: Statistical Tests of Comparative Performance - 1 Wedge Economies

Output					
Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01		A_t vs $\tau_{l,t}$	-2.62	0.01
Friedman	< 0.01	t -tests	A_t vs $\tau_{x,t}$	-4.27	< 0.01
			$\tau_{l,t}$ vs $\tau_{x,t}$	-1.76	0.09
Economy	RMSEs				
A_t	1.33		A_t vs $\tau_{l,t}$	300.00	0.04
$\tau_{l,t}$	1.92	Wilcoxon	A_t vs $\tau_{x,t}$	237.00	< 0.01
$\tau_{x,t}$	2.52		$\tau_{l,t}$ vs $\tau_{x,t}$	308.00	0.07
Hours					
Joint	p-val	Pairwise		stat	p-val
ANOVA	0.12		A_t vs $\tau_{l,t}$	-0.80	0.43
Friedman	0.08	t -tests	A_t vs $\tau_{x,t}$	-2.45	0.02
			$\tau_{l,t}$ vs $\tau_{x,t}$	-1.10	0.28
Economy	RMSEs				
A_t	1.66		A_t vs $\tau_{l,t}$	375.00	0.91
$\tau_{l,t}$	1.91	Wilcoxon	A_t vs $\tau_{x,t}$	310.00	0.08
$\tau_{x,t}$	2.33		$\tau_{l,t}$ vs $\tau_{x,t}$	322.00	0.16
Investment					
Joint	p-val	Pairwise		stat	p-val
ANOVA	0.22		A_t vs $\tau_{l,t}$	1.33	0.19
Friedman	0.03	t -tests	A_t vs $\tau_{x,t}$	-0.74	0.46
			$\tau_{l,t}$ vs $\tau_{x,t}$	-1.60	0.12
Economy	RMSEs				
A_t	9.74		A_t vs $\tau_{l,t}$	414.00	0.21
$\tau_{l,t}$	7.47	Wilcoxon	A_t vs $\tau_{x,t}$	385.00	0.68
$\tau_{x,t}$	11.80		$\tau_{l,t}$ vs $\tau_{x,t}$	335.00	0.31

Table 2: Statistical Tests of Comparative Performance - 3 Wedge Economies

Output					
Joint	p-val	Pairwise		stat	p-val
ANOVA	0.12	<i>t</i> -tests	No A_t vs No $\tau_{l,t}$	0.33	0.74
Friedman	< 0.01		No A_t vs No $\tau_{x,t}$	1.98	0.06
			No $\tau_{l,t}$ vs No $\tau_{x,t}$	1.59	0.12
Economy	RMSEs	Wilcoxon			
No A_t	2.04		No A_t vs No $\tau_{l,t}$	393.00	0.52
No $\tau_{l,t}$	1.93		No A_t vs No $\tau_{x,t}$	441.00	0.04
No $\tau_{x,t}$	1.39		No $\tau_{l,t}$ vs No $\tau_{x,t}$	438.00	0.05
Hours					
Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01	<i>t</i> -tests	No A_t vs No $\tau_{l,t}$	-5.07	< 0.01
Friedman	< 0.01		No A_t vs No $\tau_{x,t}$	-2.32	0.03
			No $\tau_{l,t}$ vs No $\tau_{x,t}$	1.98	0.06
Economy	RMSEs	Wilcoxon			
No A_t	1.07		No A_t vs No $\tau_{l,t}$	283.00	0.01
No $\tau_{l,t}$	2.59		No A_t vs No $\tau_{x,t}$	243.00	< 0.01
No $\tau_{x,t}$	1.82		No $\tau_{l,t}$ vs No $\tau_{x,t}$	403.00	0.35
Investment					
Joint	p-val	Pairwise		stat	p-val
ANOVA	0.25	<i>t</i> -tests	No A_t vs No $\tau_{l,t}$	-0.48	0.64
Friedman	0.53		No A_t vs No $\tau_{x,t}$	-1.50	0.14
			No $\tau_{l,t}$ vs No $\tau_{x,t}$	-1.13	0.26
Economy	RMSEs	Wilcoxon			
No A_t	7.68		No A_t vs No $\tau_{l,t}$	362.00	0.82
No $\tau_{l,t}$	8.68		No A_t vs No $\tau_{x,t}$	327.00	0.21
No $\tau_{x,t}$	11.89		No $\tau_{l,t}$ vs No $\tau_{x,t}$	346.00	0.48

Success Ratios for 1 Wedge and 3 Wedge Economies

Figure 13: Output

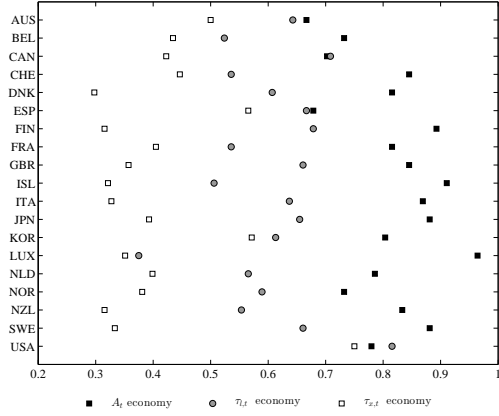


Figure 16: Output

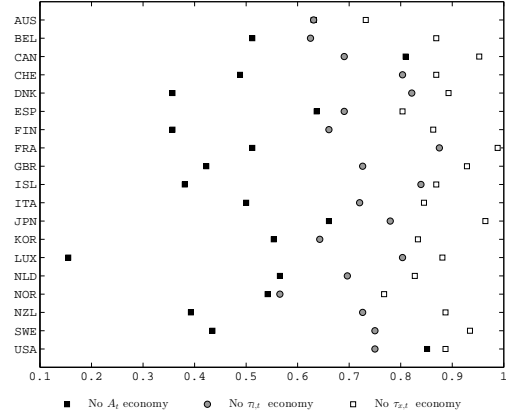


Figure 14: Hours worked

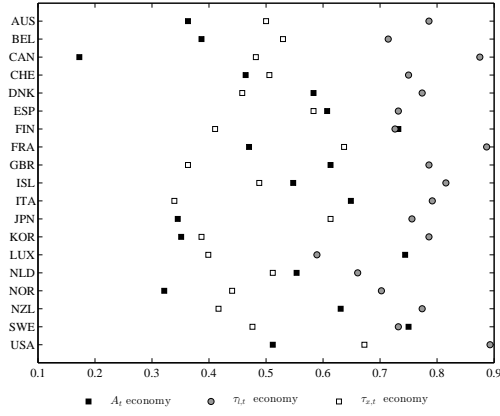


Figure 17: Hours worked

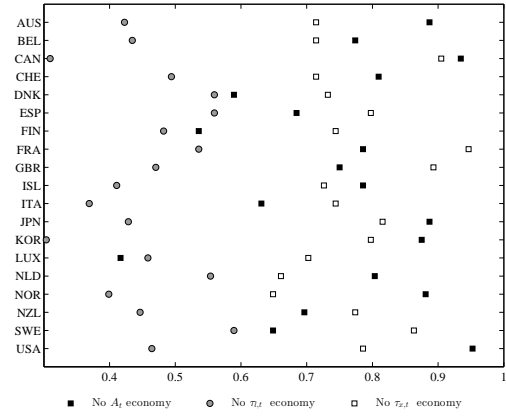


Figure 15: Investment

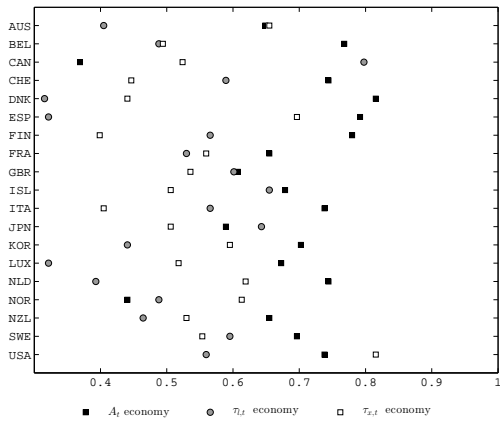
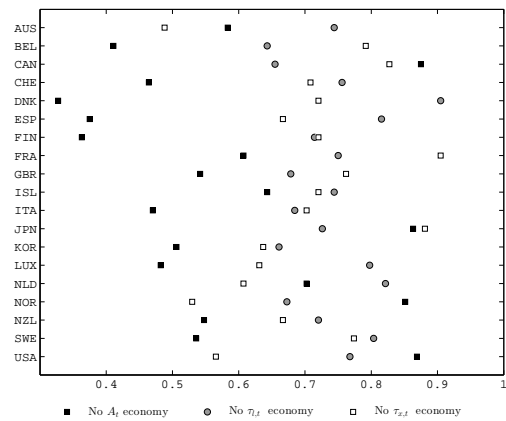


Figure 18: Investment



Statistical Tests of Comparative Performance - SRs

Table 3: Statistical Tests of Comparative Performance - 1 Wedge Economies

Output					
<hr/>					
Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01				
Friedman	< 0.01	<i>t</i> -tests	A_t vs $\tau_{l,t}$	7.19	< 0.01
			A_t vs $\tau_{x,t}$	12.34	< 0.01
			$\tau_{l,t}$ vs $\tau_{x,t}$	5.66	< 0.01
Economy	Success Ratios				
A_t	81%		A_t vs $\tau_{l,t}$	537.00	< 0.01
$\tau_{l,t}$	61%	Wilcoxon	A_t vs $\tau_{x,t}$	546.00	< 0.01
$\tau_{x,t}$	42%		$\tau_{l,t}$ vs $\tau_{x,t}$	511.50	< 0.01
<hr/>					
Hours					
<hr/>					
Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01				
Friedman	< 0.01	<i>t</i> -tests	A_t vs $\tau_{l,t}$	-6.12	< 0.01
			A_t vs $\tau_{x,t}$	0.72	0.48
			$\tau_{l,t}$ vs $\tau_{x,t}$	10.26	< 0.01
Economy	Success Ratios				
A_t	52%		A_t vs $\tau_{l,t}$	214.50	< 0.01
$\tau_{l,t}$	76%	Wilcoxon	A_t vs $\tau_{x,t}$	395.50	0.47
$\tau_{x,t}$	48%		$\tau_{l,t}$ vs $\tau_{x,t}$	547.00	< 0.01
<hr/>					
Investment					
<hr/>					
Joint	p-val	Pairwise		stat	p-val
ANOVA	< 0.01				
Friedman	< 0.01	<i>t</i> -tests	A_t vs $\tau_{l,t}$	4.15	< 0.01
			A_t vs $\tau_{x,t}$	3.62	< 0.01
			$\tau_{l,t}$ vs $\tau_{x,t}$	-0.94	0.35
Economy	Success Ratios				
A_t	68%		A_t vs $\tau_{l,t}$	497.00	< 0.01
$\tau_{l,t}$	51%	Wilcoxon	A_t vs $\tau_{x,t}$	482.50	< 0.01
$\tau_{x,t}$	55%		$\tau_{l,t}$ vs $\tau_{x,t}$	345.00	0.47

Table 4: Statistical Tests of Comparative Performance - 3 Wedge Economies

Output					
Joint		Pairwise		stat	p-val
p-val					
ANOVA	< 0.01		No A_t vs No $\tau_{l,t}$	-5.06	< 0.01
Friedman	< 0.01	t -tests	No A_t vs No $\tau_{x,t}$	-8.90	< 0.01
			No $\tau_{l,t}$ vs No $\tau_{x,t}$	-6.16	< 0.01
Economy	Success Ratios				
No A_t	51%		No A_t vs No $\tau_{l,t}$	234.50	< 0.01
No $\tau_{l,t}$	73%	Wilcoxon	No A_t vs No $\tau_{x,t}$	199.00	< 0.01
No $\tau_{x,t}$	87%		No $\tau_{l,t}$ vs No $\tau_{x,t}$	217.00	< 0.01
Hours					
Joint		Pairwise		stat	p-val
p-val					
ANOVA	< 0.01		No A_t vs No $\tau_{l,t}$	7.87	< 0.01
Friedman	< 0.01	t -tests	No A_t vs No $\tau_{x,t}$	-0.49	0.63
			No $\tau_{l,t}$ vs No $\tau_{x,t}$	-11.91	< 0.01
Economy	Success Ratios				
No A_t	75%		No A_t vs No $\tau_{l,t}$	283.00	0.01
No $\tau_{l,t}$	46%	Wilcoxon	No A_t vs No $\tau_{x,t}$	243.00	< 0.01
No $\tau_{x,t}$	77%		No $\tau_{l,t}$ vs No $\tau_{x,t}$	403.00	0.35
Investment					
Joint		Pairwise		stat	p-val
p-val					
ANOVA	< 0.01		No A_t vs No $\tau_{l,t}$	-3.66	< 0.01
Friedman	0.01	t -tests	No A_t vs No $\tau_{x,t}$	-2.50	0.02
			No $\tau_{l,t}$ vs No $\tau_{x,t}$	1.33	0.19
Economy	Success Ratios				
No A_t	58%		No A_t vs No $\tau_{l,t}$	268.50	< 0.01
No $\tau_{l,t}$	74%	Wilcoxon	No A_t vs No $\tau_{x,t}$	286.00	0.01
No $\tau_{x,t}$	70%		No $\tau_{l,t}$ vs No $\tau_{x,t}$	415.00	0.20