Lindé, Jesper

Working Paper
The Effects of Permanent Technology Shocks on Labor Productivity and Hours in the RBC model

Sveriges Riksbank Working Paper Series, No. 161

Provided in Cooperation with:
Central Bank of Sweden, Stockholm

Suggested Citation: Lindé, Jesper (2004) : The Effects of Permanent Technology Shocks on Labor Productivity and Hours in the RBC model, Sveriges Riksbank Working Paper Series, No. 161, Sveriges Riksbank, Stockholm

This Version is available at:
http://hdl.handle.net/10419/82490

Terms of use:
Documents in EconStor may be saved and copied for your personal and scholarly purposes.
You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.
The Effects of Permanent Technology Shocks on Labor Productivity and Hours in the RBC model

Jesper Lindé

APRIL 2004
WORKING PAPERS ARE OBTAINABLE FROM

Sveriges Riksbank • Information Riksbank • SE-103 37 Stockholm
Fax international: +46 8 787 05 26
Telephone international: +46 8 787 01 00
E-mail: info@riksbank.se

The Working Paper series presents reports on matters in the sphere of activities of the Riksbank that are considered to be of interest to a wider public. The papers are to be regarded as reports on ongoing studies and the authors will be pleased to receive comments.

The views expressed in Working Papers are solely the responsibility of the authors and should not to be interpreted as reflecting the views of the Executive Board of Sveriges Riksbank.
The Effects of Permanent Technology Shocks on Labor Productivity and Hours in the RBC model

Jesper Lindé*

Sveriges Riksbank Working Paper Series
No. 161
April 2004

Abstract

Recent work on the effects of permanent technology shocks argue that the basic RBC model cannot account for a negative correlation between hours worked and labor productivity. In this paper, I show that this conjecture is not necessarily correct. In the basic RBC model, I find that hours worked fall and labor productivity rises after a positive permanent technology shock once one allows for the possibility that the process for the permanent technology shock is slightly persistent in growth rates. A more serious limitation of the RBC model is its inability to generate a persistent rise in hours worked after a positive permanent technology shock along with a rise in labor productivity that are in line with what the data suggests. These results call for a reconsideration of the real and nominal frictions and policy response that need to be introduced in the basic RBC model in order to improve the model's ability to match the data.

Keywords: permanent technology shocks; hours worked per capita; labor productivity; real business cycle model; vector autoregressions.

JEL Classification Numbers: E32, E24.

*Research Department, Sveriges Riksbank, 103 37 Stockholm, Sweden. E-mail: jesper.linde@riksbank.se. I am grateful for helpful discussions and comments by Susanto Basu, Lawrence Christiano, Jordi Galí, Tor Jacobson, Marianne Nessén and seminar participants at Sveriges Riksbank and Stockholm School of Economics. The views expressed in this paper are solely the responsibility of the author and should not be interpreted as reflecting the views of the Executive Board of Sveriges Riksbank.
1 Introduction

Using the identifying assumption that permanent technology shocks are the only shocks that have an effect on the long-run level of productivity, Galí (1999) reports that hours worked fall while labor productivity rises after a positive permanent shock to technology. Based on this empirical result, Galí argues that the standard Real Business Cycle (RBC) model is at odds with the data for two reasons. First, he argues that the standard RBC model implies that hours worked rise after a positive permanent innovation to technology, whereas the response is negative in the data according to his estimates. Second, because he provides evidence that hours worked and labor productivity are actually strongly positively correlated in the data, while the permanent technology shock he identifies is a source of negative correlation among these variables, he concludes that some other shock(s) must be the driving source of business cycles.

Not surprisingly, the seminal work of Galí (1999) has generated considerable interest in the effects of permanent technology shocks. One set of papers, see Basu, Fernald and Kimball (2001), Francis and Ramey (2001), and Galí (2003), have obtained similar empirical results. Therefore, these papers have tried to introduce modifications to the standard RBC model in order to account for negative correlation of labor productivity and per capita hours worked induced by permanent technology shocks. Another set of papers challenges Galí’s empirical results and argues that both labor productivity and hours worked actually rise after a positive permanent technology innovation, see Altig et al. (2002) and Christiano, Eichenbaum and Vigfusson (2003a, 2003b).

In this paper, I take one step back and study the effects of permanent technology shocks in the “plain-vanilla” RBC model with everything being completely standard. To gain some insight about how the permanent technology shock works in the model, I also include transitory technology shocks for comparison. As in Altig et al. (2002), I allow for the possibility that persistent processes characterize both permanent and temporary technology shocks.

In contrast to the conjectures in the above mentioned literature, I find using a completely standard calibration of the basic RBC model, that a positive permanent technology shock implies that hours worked may fall substantially while labor productivity increases. The reason why this can happen is that the process for the permanent technology is slightly persistent in growth rates, thus making it profitable to decrease the supply of hours until labor effort becomes most effective.1 Moreover, by employing the same identifying assumption as in Galí (1999) I show

---

1 It should be stressed that other papers have also emphasized the possibility that hours fall while labor productivity rise after a positive permanent technology shock, but these papers do not use the standard RBC model, or the assumption of perfect information about the shock realization as I do. An example of the former type is the recent paper by Rotemberg (2003), who use a model with $N_t$ types of capital in service at time $t$ where each type of capital is associated with a different technology parameter $z_j$. Rotemberg (2003) shows that if the permanent technology shock diffuses slowly in the economy (i.e. is correlated in growth rates and does not reach the new steady state level contemporaneously), hours fall while labor productivity rise after a positive
that the basic RBC model is able to match the estimated impulse response functions for a broad set of variables well when hours worked per capita are treated as difference stationary (as in Galí, 1999) in the data. However, when hours worked are treated as level stationary in the empirical analysis, as in Christiano, Eichenbaum and Vigfusson (2003a, 2003b), the RBC model cannot account for the resulting strong and persistent rise in hours worked along with a moderate response of labor productivity, consumption, investment and the real wage. Thus, in contrast to the conclusions in the previous literature, these results suggest that the RBC model is actually able to replicate a substantial fall in hours and a rise in labor productivity reported by Galí (1999), Basu, Fernald and Kimball (2001), and Francis and Ramey (2001), while it cannot replicate the persistent rise in hours and labor productivity reported by Christiano, Eichenbaum and Vigfusson (2003a, 2003b). Since Christiano, Eichenbaum and Vigfusson (2003a) argue convincingly that hours worked are level stationary rather than difference stationary and thus that hours rise persistently after a positive technology shock, the inability of the basic RBC model to replicate this result calls for a reconsideration of how the basic RBC model needs to be modified in order to with the data. I elaborate further on this last issue in the conclusions.²

The paper is structured as follows. In the next section, I briefly present the standard RBC model that is used in the paper. Sections 2 and 3 show the effects of transitory and permanent technology shocks in the RBC model. In Section 4, I investigate the model’s ability to reproduce some estimated impulse response functions in the data. Finally, some conclusions are provided in Section 5.

2 The standard RBC model with temporary and permanent technology shocks

Consider a representative agent which maximizes expected utility with preferences summarized by

\[ E_0 \sum_{t=0}^{\infty} \beta^t u(C_t, H_t), \]

subject to the following inter-temporal budget constraint

\[ C_t + I_t = W_t H_t + R_t^K K_t. \]

Investment in period \( t \) produce productive capital in period \( t + 1 \) according to

\[ K_{t+1} = (1 - \delta) K_t + I_t, \]

2 Christiano, Eichenbaum and Vigfusson (2003a) provide evidence by means of multivariate KPSS and ADF tests (see Hansen, 1995) - which they demonstrate have substantially more power than standard univariate tests - that hours worked per capita is stationary. This empirical result is appealing because in most theoretical models, hours worked per capita is stationary.
where $\delta$ is the rate of capital depreciation.

The representative firm has access to the following production technology

$$Y_t = z_t^{1-\theta} e^{\varepsilon_t} K_t^\theta H_t^{1-\theta},$$  \hspace{1cm} (4)

where the processes for $z_t$ and $\varepsilon_t$ are given by

$$\ln z_t = \ln z_{t-1} + x_t,$$

$$x_t = (1 - \rho_x) \mu_z + \rho_x x_{t-1} + \eta_t, 0 < \rho_x < 1, \eta_t \sim N(0, \sigma_{\eta}^2),$$

$$\varepsilon_t = \rho_{\varepsilon}\varepsilon_{t-1} + \nu_t, 0 < \rho_{\varepsilon} < 1, \nu_t \sim N(0, \sigma_{\nu}^2).$$

Consequently, $\eta_t$ is a permanent technology shock and $\nu_t$ is a temporary technology shock. $\eta_t$ is a permanent shock in the sense that it will have a permanent effect on labor productivity ($Y_t/H_t$) in the long run, whereas $\nu_t$ will not. Since $H_t$, the share of available time spent in employment, is stationary, this implies that both $Y_t$ and $Y_t/H_t$ will rise permanently after a positive realization of $\eta_t$, whereas $Y_t$ and $Y_t/H_t$ will return to their steady-state values after a positive realization of $\nu_t$. Finally, note that the steady-state growth rate in this economy is given by $\mu_z$.

If we assume that factor markets are characterized by perfect competition, the real pre-depreciation rental rate on capital, $R_t^K$, and the real wage $W_t$ are given by

$$R_t^K = \theta z_t^{1-\theta} e^{\varepsilon_t} \left( \frac{K_t}{H_t} \right)^{\theta-1},$$

$$W_t = (1 - \theta) z_t^{1-\theta} e^{\varepsilon_t} \left( \frac{K_t}{H_t} \right)^{\theta}. \hspace{1cm} (6)$$

Using (6) and (3) in (2), we have

$$C_t + K_{t+1} = z_t^{1-\theta} e^{\varepsilon_t} K_t^\theta H_t^{1-\theta} + (1 - \delta) K_t. \hspace{1cm} (7)$$

The dynamic program for the representative individual in this economy can be written as

$$V(K_t) \equiv \max_{\{C_t, H_t\}} \{u(C_t, H_t) + E_t \beta V(K_{t+1})\}$$

s.t. (7)

where we notice that off-work time $L_t$ and work time $H_t$ have been normalized to sum to unity.

In order to compute the steady state and the decision rules in this economy, we need to specify the functional form for $u(C_t, 1 - H_t)$. As a benchmark, we use the standard functional form when consumption and leisure are non-separable and given by

$$u(C_t, H_t) = \frac{C_t^\sigma (1 - H_t)^{1-\sigma}}{1 - \sigma} - 1. \hspace{1cm} (9)$$
This is perhaps the most commonly used specification in the RBC-literature, see e.g. Hansen and Prescott (1995).\footnote{I have also checked the robustness of the results reported in this paper when the utility function is separable in consumption and leisure (see e.g. Hansen, 1995), i.e. \( u(C_t, H_t) = \ln C_t - \lambda H_t \), where \( \lambda \) is calibrated so that hours worked in steady state equals the steady-state value for \( H \) in the benchmark model (non-separable utility function). The main results of the paper are completely invariant to this specification.}

Since the technology shock \( z_t \) is permanent, we need to scale all variables except \( H_t \) with this shock in order to be able to solve for a constant steady state and to compute the decision rules for the representative agent in this economy. Let \( \hat{C}_t \equiv C_t / z_t, \hat{Y}_t \equiv Y_t / z_t, \hat{I}_t \equiv I_t / z_t \) and \( \hat{K}_{t+1} \equiv K_{t+1} / z_t \). As shown in Appendix A, we then have the following system of equations that characterize this economy

\[
\hat{C}_t (1 - \alpha) = \alpha (1 - H_t) (1 - \theta) x_t^{-\theta} e^{\varepsilon_t} \left( \frac{\hat{K}_t}{H_t} \right)^\theta, \\
\hat{C}_t^{\alpha(1-\sigma)^{-1}} (1 - H_t)^{(1-\alpha)(1-\sigma)} = \beta E_t \left[ x_{t+1}^{\alpha(1-\sigma)^{-1}} \left( \hat{C}_{t+1}^{\alpha(1-\sigma)^{-1}} (1 - H_{t+1})^{(1-\alpha)(1-\sigma)} \right) \times \left( 1 + x_{t+1}^{1-\theta} e^{\varepsilon_{t+1}} \left( \frac{\hat{K}_{t+1}}{H_{t+1}} \right)^{\theta-1} - \delta \right) \right], \\
\hat{C}_t + \hat{I}_t = x_t^{-\theta} e^{\varepsilon_t} \hat{K}_t^{\theta} H_t^{1-\theta}, \\
\hat{K}_{t+1} = (1 - \delta) \hat{K}_t x_t^{-1} + \hat{I}_t, \\
\hat{Y}_t = \hat{C}_t + \hat{I}_t.
\]

Together with (5), the set of equations in (10) constitute the model that we want to solve. Note that we scale \( K_{t+1} \) with \( z_t \) rather than \( z_{t+1} \), because \( K_{t+1} \) is determined in period \( t \). Appendix A contains the details for the solution of the model.

To calibrate the model, I use standard parameters in the RBC-literature. More specifically, the model is calibrated to match quarterly data by setting \( \beta = 0.99, \theta = 0.36, \delta = 0.025, \alpha = 0.33, \sigma = 2 \), and the steady-state growth rate in the model, \( \mu_x \), to 1.03\footnote{Note that these two parameters along with the inverse of the intertemporal elasticity of substitution (\( \sigma \)) are estimated in Section 5.}. These values are standard in the literature, see e.g. Cooley and Prescott (1995) and Christiano, Eichenbaum and Evans (2001). To calibrate the exogenous processes for the stationary technology shock (\( \varepsilon_t \)) and the unit-root technology shock (\( \hat{x}_t \)), I use the estimates in Altig et al. (2002) and set \( \rho_x = 0.80 \) and \( \sigma_x = 0.11 \). In order to highlight the difference in propagation of unit-root and correlated stationary technology shocks, I use \( \rho_x = 0.80 \) and \( \sigma_v = 0.11 \) as well.

3 Impulse response functions to a temporary technology shock

The impulse response functions to a temporary shock is shown in Figure 1. I include the effects on six variables in the figure; output, consumption, investment, hours worked, labor productivity...
(defined as $\hat{y}_t + \ln z_t - h_t$), and the level of technology (defined as $\ln z_t + \epsilon_t$). As is standard in the literature, we see that hours, output and labor productivity all go up after a positive temporary technology. Thus, as can be seen from Figure 1, if this type of shock is the key driving force of business cycles, we ought to see a strong positive correlation between hours worked and labor productivity in the data.

4 Impulse response functions to a permanent technology shock

The solid line in Figure 2 shows the impulse response functions for a permanent technology shock for the baseline calibration of the shock process. Perhaps surprisingly, we see that hours worked fall for the first 5 quarters after the shock, in contrast to the conjectures made by Galí (1999) and Christiano, Eichenbaum and Vigfusson (2003a). Moreover, it can be seen that labor productivity goes up despite the fall in hours worked, inducing a negative correlation between hours worked and labor productivity, a feature of the empirical results which Galí (1999) emphasized was strong evidence against the RBC model. It is also notable that investment drop in the first quarters after the shock.

The question then arrises what the underlying mechanism is in the RBC model that enables it to generate a negative short-run response in hours worked while labor productivity goes up. To understand why, the dashed line in Figure 2 shows the impulse response functions to a zero correlated shock in growth rates (i.e. using $\rho_x = 0$) where I set the standard deviation of this shock so that the long-term impact on the level of technology shock is identical to the growth rate correlated shock. In this case, we see that the conjecture by Galí (1999) and Christiano, Eichenbaum and Vigfusson (2003a) holds; hours worked go up, and so do labor productivity. Why do the results for hours and investment differ when the permanent shock is persistent in growth rates? My intuition is that when the permanent technology shock is persistent in growth rates, it is more profitable for individuals to work and invest less as the shock hits the economy because it takes a while before labor and capital input become most productive (compare lines for technology in Figure 2), whereas in the case when the technology shock is non-permanent or non-persistent, labor and capital input rise because they become most productive immediately. Therefore, we obtain a fall in hours and investment after a positive permanent technology shock if the shock is slightly correlated in growth terms. However, due to the permanent income mechanism (i.e. the wealth effect stemming from the permanent improvement in technology), consumption rises immediately and thus labor productivity too irrespective of the shock’s persistence, inducing a negative correlation between hours worked and labor productivity when the shock is correlated in growth terms.

---

5 Lower case variables indicate natural logarithms. Also, note that $\hat{y}_t + \ln z_t$ is output in levels, whereas $\hat{y}_t$ is the “detrended” output. Same transformation have been applied for consumption, investment and the real wage.

6 Interestingly, Basu, Fernald, and Kimball (2001) argue that the RBC model is in no case able to generate a
in Figure 2 suggest that the consumption response after a positive permanent technology is potentially a more informative way to check the consistency of the RBC model with the data compared to the response of hours worked. No matter the process for the permanent technology shock, consumption should always rise after a positive technology shock, in contrast to hours worked.

Some other aspects of the parameterization of RBC model should be emphasized. The results are not very sensitive to the choice of $\sigma$ (the inverse of the inter-temporal substitution elasticity) between 1 (log-utility) and setting $\sigma = 10$ (which is a fairly high number). Moreover, as might be evident from Figure 2, a larger value for the persistence parameter $\rho_x$ than estimated by Altig et al. (2002) only increases the fall in hours and the negative correlation between labor productivity and hours worked in the RBC model. It should be emphasized that only a very small value for the persistence parameter $\rho_x$ is required in order to generate a substantial fall in hours worked. Finally, using the Hansen (1985) type utility function, where consumption and hours are additive do not change the results either, such a utility function rather amplifies the mechanism since consumption and hours are additive in the utility function.

5 Taking the basic RBC model to the data

In this section, I will examine if the basic RBC model can replicate the impulse response functions in the data by matching the impulse response functions in the model with the ones obtained in a vector autoregressive (VAR) model using the same identifying assumptions as by Galí (1999), i.e. that the permanent technology shock is the only source of fluctuations in labor productivity in the long run. We will vary the parameters $\sigma$, $\rho_x$, and $\sigma_\eta$ as to minimize the following criterion

$$J = \left( \hat{\psi} - \psi(\sigma, \rho_x, \sigma_\eta) \right)' \hat{V}^{-1} \left( \hat{\psi} - \psi(\sigma, \rho_x, \sigma_\eta) \right)$$

where $\hat{\psi}$ is a $(j \times k)$ vector which contains the impulse response functions $j$ periods following the shock for the $k$ variables of interest in the data, whereas $\psi(\rho_x, \sigma_\eta)$ is a $(j \times k)$ vector with the corresponding impulse response functions in the model. $\hat{V}$ is a diagonal matrix with the estimated standard deviation of each response in the data. I set $j = 20$ in order to study the RBC-models’ properties in both the short and long run. The procedure adopted is exactly the same as the one used by Christiano, Eichenbaum and Evans (2001), and Altig et al. (2002).

To generate the impulse response functions in the data, we use two VAR models. First,
I will estimate a bivariate VAR containing labor productivity and hours worked. Second, I estimate a VAR augmented by consumption, investment and the real wage. Since the recent empirical literature have stressed the importance of whether hours worked are first differenced or not, I will consider two cases. The first case will be when hours worked are first differenced in the data (as in Galí 1999, 2003), although hours worked is stationary in the model. Second, since Christiano, Eichenbaum and Vigfusson (2003a) argue that hours worked should not be differenced, I also present results for the VARs estimated with hours worked in levels. In the former case, we anticipate that hours worked will fall and labor productivity will rise whereas in the latter case we expect that both hours worked and labor productivity will rise after a positive technology shock. The variables used in the VARs are depicted in Figure 3.

In Figure 4, I show the results when matching the models impulse response functions with the ones obtained in the estimated bivariate VAR with hours in first differences. The parameter estimates of $\sigma$, $\rho$, and $\sigma_\eta$ for this VAR are reported in Table 1, along with the estimates for the other VAR models. We see that although the model is able to generate a substantial drop in hours worked per capita ($-0.2$ percent) and rise in labor productivity, the response of hours worked is still slightly outside the grey area (which indicates the 95-percent confidence interval) in the first period. It should be pointed out that if $\sigma$ is restricted to 2 (i.e. the intertemporal elasticity of substitution is considerably higher than estimated), then the RBC model still produces a sizeable initial fall in $H_t$ of $-0.12$ percent. So as noted by Rotemberg and Woodford (1996), the estimated low intertemporal elasticity of substitution is an important ingredient why the model is able to generate a substantial fall in $H_t$. However, allowing the permanent technology shock to be correlated in growth rates is a more important channel. If I restrict the permanent technology shock to a random walk (i.e. non-correlated in growth terms),

7 See Altig et al. (2002) for details about the data, the sample period is 1959Q1 – 2001Q4. 4 lags are used in the estimations. When estimating the VAR, I impose the restriction that the permanent technology shock is the only shock that influence labor productivity in the long run. Moreover, in the VARs estimated with hours in first differences, I estimate the hours equation in double differences for hours (using first differences of labor productivity and hours as instruments) in order to impose the restriction that hours return to nil after a permanent technology shock. The standard errors in $V$ are bootstrapped (using 1000 repetitions).

8 I follow Altig et al. (2002) and include $\ln(C_t/Y_t)$, $\ln(I_t/Y_t)$, and $\ln(W_t/Y_t)$ as variables and convert the ratios to levels when computing the impulse response functions. These variables are stationary in the model and also appear to be stationary in the data (see Figure 3).

9 Since the modulus of the largest root of the characteristic polynomial was very close to unity (0.993) for the VAR with hours in levels (implying that the bootstrapped standard errors bands for the impulse response functions become very wide), I expanded the sample for the VAR with hours in levels by starting 1948Q1 instead of 1960Q1. By this procedure, the modulus of the largest root was considerably lower (0.94). As an alternative, I also detrended hours worked with a linear-quadratic trend prior to estimation of the VARs with hours in levels for the benchmark sample period. For the 5-variables VAR, I then obtained very similar results as the ones reported in Figure 7. However, the response of hours in a bivariate VAR is very sensitive to whether hours worked are detrended or not. If hours are detrended, hours fall initially consistent with the findings by Galí (1999, 2003), but if hours are not detrended hours rise strongly as can be seen in Figure 6. Thus, the empirical finding that hours fall even when it is detrended with a linear-quadratic trend appear to be sensitive to the specification of the VAR model. This mechanism is also evident by comparing the empirical impulse response functions in Figures 4 and 5. In the bivariate VAR for hours in first differences hours fall strongly, whereas in the 5-variables VAR with hours in first differences, there is only a minor initial fall followed by an increase in hours.

10 It should be pointed out that the model is able to match the impulse response functions of $H_t$ only very well, but then the response of labor productivity is too strong relative to the data.
the initial fall in hours is only −0.07 percent.

However, for the VAR with hours in first differences with the larger set of variables, the RBC model fits quite nicely as can be seen in Figure 5. In this case, the estimated initial drop in the data for hours worked is not as pronounced as in the bivariate VAR and the model responses are well inside the grey area and close to the empirical point estimates for all variables in both the short and long run. The parameter estimates of $\sigma$, $\rho_x$, and $\sigma_\eta$ for this VAR are all very reasonable. For $\sigma$, the number could easily have been the outcome of an ordinary calibration procedure, and for $\rho_x$ it is interesting to note that we do not need much persistence in the permanent shock in growth rates in order to line up the model with the data.

### Table 1: Parameter estimates when matching the RBC model with US data.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Bivariate VAR model</th>
<th>5-variables VAR model</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\rho_x$</td>
<td>0.43</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma_\eta$</td>
<td>0.55</td>
<td>0.43</td>
</tr>
<tr>
<td>$\sigma$</td>
<td>7.28</td>
<td>0.75</td>
</tr>
</tbody>
</table>

Turning to the bivariate VAR with hours worked per capita in levels in Figure 6, we first notice that we obtain a “hump-shaped” rise in hours worked instead of a fall as in Figure 4. The short-run response of labor productivity is also somewhat stronger than in Figure 4, but the long-run response of labor productivity is about the same. However, as was the case in Figure 4, the RBC model is not able to replicate the estimated joint effects of the technology shock in the data, the initial effect on labor productivity is too weak and the hours response is also too weak compared to the data although it is inside the gray area. The parameter estimates of $\sigma$, $\rho_x$ and $\sigma_\eta$ for this VAR model are quite different than for the VAR model with hours in first differences. Given that we obtain a rise in hours worked, it comes as no surprise that the estimated value of $\rho_x$ is zero, whereas the estimate of $\sigma$ is very low. Figure 7 presents the results for the 5-variables VAR with hours in levels. As can be seen from the figure, the RBC model cannot match the obtained impulse response functions in the data for hours worked with this specification because hours worked rise strongly after the shock in line with the results in Christiano, Eichenbaum and Vigfusson (2003a). The parameter estimates of $\sigma$, $\rho_x$, and $\sigma_\eta$ for this VAR are 0.01 (corner solution), 0.01, and 0.57, respectively. The estimate of $\sigma$ is implausibly low, implying that the utility function is almost linear. Imposing $\sigma = 2$ instead, we obtain an even lower response of hours.

To sum up, it is clear by comparison of Figures 5 and 7 reveal that the RBC model can easier accommodate a fall in hours worked along with a rise in labor productivity after a positive permanent technology shock than vice versa, in contrast to the conjectures in the empirical
literature, see e.g. Galí (1999), Basu, Fernald and Kimball (2001), Christiano, Eichenbaum and Vigfusson (2003a), and Francis and Ramey (2001). Moreover, the estimated persistence of the technology shock in growth rates to achieve these results is very low, around 0.3 – 0.4 at the quarterly frequency and thus less than 0.05 at an annual frequency.

6 Conclusions

In this paper, I have shown that the standard RBC model (Cooley and Prescott, 1995) can produce a substantial fall in hours worked along with a reasonable rise in labor productivity after a positive permanent technology shock, once one allows for the possibility that the technology shock is slightly correlated over time in growth terms.

This finding implies that the evidence presented by e.g. Galí (1999), i.e. that hours worked drop while labor productivity rises after a permanent technology shock, cannot be taken as evidence against the RBC model of US business cycles. Moreover, and very importantly, this also implies that it is harder to argue that the empirical results provide indirect evidence that we need sticky prices along with non-accommodative monetary policy (Basu, Fernald and Kimball, 2001, and Galí, 1999) or real frictions (Francis and Ramey, 2001) in the model in order to generate a fall in hours worked.

However, I have also shown that the basic RBC model cannot accommodate the persistent rise in hours worked following a permanent technology shock reported by Altig et al. (2002) and Christiano, Eichenbaum and Vigfusson (2003a) when hours worked per capita are treated as level stationary in the data. According to their estimates (similar results were obtained in this paper in a 5-variables VAR), the response of hours is “hump-shaped” with a peak effect after 1 – 2 years.\footnote{Fernald (2003) argues using bivariate VARs that by allowing for breaks in mean labor productivity growth in 1973 and 1995, the empirical findings in Galí/Francis-Ramey and Christiano, Eichenbaum and Vigfusson (2003a) can be reconciled and hours fall about –0.2 percent on impact after a positive technology shock irrespectively if hours are included in levels or growth rates in the VAR. Although further analysis is needed before we may draw further firmer conclusions regarding this finding, we notice here that if Fernald is correct, then the conclusions made in this paper regarding the frictions and monetary policy response that we need to bring in to the RBC model in order to make it consistent with the dynamic effects of technology shocks will be altered.} In order to account for this “hump-shaped” rise, some real and nominal frictions along with monetary policy accommodation need to be incorporated in the basic RBC model. Altig et al. (2002) present an example of one such model economy. The important difference in this model compared to the modifications of the RBC model suggested by Galí (1999), Basu, Fernald and Kimball (2001) and Francis and Ramey (2001) being that it is accommodative monetary policy along with sticky wages and real frictions which enables the model to replicate
the rise in hours worked and labor productivity after a positive permanent technology shock.12

Moreover, the observation that the identified neutral permanent technology shock need not be the major source of business cycle fluctuations in the data cannot be used as evidence against the notion of technology driven business cycles in a modified version of the RBC model.13 Although the data suggests that the identified permanent neutral technology shock is perhaps not the most important shock for business cycles, it is possible that other types of technology shocks can account for most of the cyclical variation. For example, we have seen in this paper (see Figure 1) that a temporary technology shock induces a strong positive correlation between hours worked and labor productivity.14 Another possibility is that investment-specific shocks account for more of the business cycle fluctuations. Using identified VARs, Fisher (2002) reports that permanent investment-specific technology shocks seem to be an important source of US business cycles.

Finally, one word of caution is that some recent papers question the ability of the identified VAR approach employed in this literature to correctly identify the shocks in the short run on sample sizes that are empirically relevant, see e.g. Erceg, Guerrieri and Gust (2003), Rotemberg (2003) and Uhlig (2003). The message from these papers is that the inability of the RBC model to correctly reproduce the estimated impulse response functions in the data could in some instances be related to an inappropriate identification of the shocks in the data.

12 Note that the “hump-shaped” rise in hours worked cannot be replicated with real and nominal frictions unless monetary policy is accommodative. The reason for this is that (i) real frictions (e.g. investment adjustment costs and habit persistence in consumer preferences) will tend to decrease hours worked even more initially (see e.g. Francis and Ramey, 2001), and (ii) nominal frictions (e.g. sticky nominal prices and wages) will also tend to drive down the hours response even more because prices and wages are expected to rise more in the future than initially. Therefore, monetary policy must be strongly accommodative in order to generate a substantial persistent rise in hours worked after a positive technology shock.

13 It should be emphasized that the share of output fluctuations due to the identified technology shock are dependent on (i) whether hours worked are treated as difference or level stationary in the VARs, and (ii) the set of variables that are included in the VARs. If hours worked are treated as level stationary, then the identified permanent technology shock account for a substantial share of the business cycle fluctuations (around 40 percent in the 5-variables VAR), whereas if hours worked are treated as difference stationary it accounts for a small share of the business cycle fluctuations.

14 Basu, Fernald and Kimball (2001), henceforth BFK, compute a direct measure of technology that builds on Solow residual accounting and report that hours worked fall after a positive innovation in technology. Since their technology series is a mixture of permanent and transitory technology shocks, this evidence suggest that neither permanent nor temporary disembodied technology shocks are an important source of US business cycles. However, Christiano, Eichenbaum and Vigfusson (2003b), using the BFK technology series, show that the BFK results can be reversed (i.e. CEV obtain a strong rise in hours worked) by allowing for measurement errors and including hours in levels instead of first differenced hours in the analysis. Thus, we cannot rule out the possibility that transitory and/or permanent disembodied technology shocks are an important source of business cycles.
References


Erceg, Christopher, Luca Guerrieri, and Christopher Gust, (2003) “Can Long-Run Restrictions Identify Technology Shocks?”, manuscript (October 2003), Board of Governors.


Figure 1: Impulse response functions to a temporary technology shock. Basline parameterization.
Figure 2: Impulse response functions to a permanent technology shock. Baseline parameterization (solid line) and when the persistence parameter ($\rho_x$) equals 0 (dashed line).
Figure 3: A plot of the U.S. data used in the VARs.
Figure 4: Impulse response functions in the RBC-model (solid line) and the data (line with +) to a permanent technology shock. Bivariate VAR estimated for hours in first differences. Grey area indicates 95-percent confidence interval.
Figure 5: Impulse response functions in the RBC-model (solid line) and the data (line with +) to a permanent technology shock. 5-variables VAR estimated for hours in first differences. Grey area indicates 95-percent confidence interval.
Figure 6: Impulse response functions in the RBC-model (solid line) and the data (line with +) to a permanent technology shock. Bivariate VAR estimated with hours in levels. Grey area indicates 95-percent confidence interval.
Figure 7: Impulse response functions in the RBC-model (solid line) and the data (line with +) to a permanent technology shock. 5-variables VAR estimated with hours in levels. Grey area indicates 95-percent confidence interval.
Earlier Working Papers:

Inflation Forecast Targeting: the Swedish Experience by Claes Berg ........................................ 2000:100
Wage Effects of Mobility, Unemployment Benefits and Benefit Financing by Hans Lindblad ........................................ 2000:101
A Bivariate Distribution for Inflation and Output Forecasts by Mårten Blix and Peter Sellin ........................................ 2000:102
Optimal Horizons for Inflation Targeting by Nicoletta Batini and Edward Nelson ........................................ 2000:103
Price-level targeting versus inflation targeting in a forward-looking model by David Vestin ........................................ 2000:106
Unemployment and Inflation Regimes by Anders Vredin and Anders Warne ........................................ 2000:107
An Expectations-Augmented Phillips Curve in an Open Economy by Kerstin Hallsten ........................................ 2000:108
Core inflation and monetary policy by Marianne Nessén and Ulf Söderström ........................................ 2000:110
Estimating the Implied Distribution of the Future Short-Term Interest Rate Using the Longstaff-Schwartz Model by Peter Hördahl ........................................ 2000:111
Financial Variables and the Conduct of Monetary Policy by Charles Goodhart and Boris Hofmann ........................................ 2000:112
Testing for the Lucas Critique: A Quantitative Investigation by Jesper Lindé ........................................ 2000:113
Monetary Policy Analysis in Backward-Looking Models by Jesper Lindé ........................................ 2000:114
UIP for short investments in long-term bonds by Annika Alexius ........................................ 2000:115
Qualitative Survey Responses and Production over the Business Cycle by Tomas Lindström ........................................ 2000:116
Supply stocks and real exchange rates by Annika Alexius ........................................ 2000:117
Casualty and Regime Inference in a Markov Switching VAR by Anders Warne ........................................ 2000:118
Average Inflation Targeting by Marianne Nessén and David Vestin ........................................ 2000:119
What have we learned from empirical tests of the monetary transmission effect? by Stefan Norbin ........................................ 2000:121
Simple monetary policy rules and exchange rate uncertainty by Kai Leitemo and Ulf Söderström ........................................ 2001:122
Targeting inflation with a prominent role for money by Ulf Söderström ........................................ 2001:123
An Alternative Explanation of the Price Puzzle by Paolo Giordani ........................................ 2001:125
Interoperability and Network Externalities in Electronic Payments by Gabriela Guibourg ........................................ 2001:126
Monetary Policy with Incomplete Exchange Rate Pass-Through by Malin Adolfson ........................................ 2001:127
Diversification and Delegation in Firms by Vittoria Cerasi and Sonja Daltung ........................................ 2001:131
Monetary Policy Signaling and Movements in the Swedish Term Structure of Interest Rates by Malin Andersson, Hans Dillén and Peter Sellin ........................................ 2001:132
Evaluation of exchange rate forecasts for the krona’s nominal effective exchange rate by Henrik Degrér, Jan Hansen and Peter Sellin ........................................ 2001:133
Identifying the Effects of Monetary Policy Shocks in an Open Economy by Tor Jacobsson, Per Jansson, Anders Vredin and Anders Warne ........................................ 2002:134
<table>
<thead>
<tr>
<th>Title</th>
<th>Authors</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>Implications of Exchange Rate Objectives under Incomplete Exchange Rate Pass-Through</td>
<td>Malin Adolfson</td>
<td>2002:135</td>
</tr>
<tr>
<td>Finding Good Predictors for Inflation: A Bayesian Model Averaging Approach</td>
<td>Tor Jacobson and Sune Karlsson</td>
<td>2002:138</td>
</tr>
<tr>
<td>How Important Is Precommitment for Monetary Policy?</td>
<td>Richard Dennis and Ulf Söderström</td>
<td>2002:139</td>
</tr>
<tr>
<td>Inflation Targeting and the Dynamics of the Transmission Mechanism</td>
<td>Hans Dillén</td>
<td>2002:141</td>
</tr>
<tr>
<td>Capital Charges under Basel II: Corporate Credit Risk Modelling and the Macro Economy</td>
<td>Kenneth Carling, Tor Jacobson, Jesper Lindé and Kasper Roszbach</td>
<td>2002:142</td>
</tr>
<tr>
<td>Capital Adjustment Patterns in Swedish Manufacturing Firms: What Model Do They Suggest?</td>
<td>Mikael Carlsson and Stefan Laséen</td>
<td>2002:143</td>
</tr>
<tr>
<td>Bank Lending, Geographical Distance, and Credit risk: An Empirical Assessment of the Church Tower Principle</td>
<td>Kenneth Carling and Sofia Lundberg</td>
<td>2002:144</td>
</tr>
<tr>
<td>Inflation, Exchange Rates and PPP in a Multivariate Panel Cointegration Model</td>
<td>Tor Jacobson, Johan Lyhagen, Rolf Larsson and Marianne Nessén</td>
<td>2002:145</td>
</tr>
<tr>
<td>Evaluating Implied RNDs by some New Confidence Interval Estimation Techniques</td>
<td>Magnus Andersson and Magnus Lomakka</td>
<td>2003:146</td>
</tr>
<tr>
<td>Taylor Rules and the Predictability of Interest Rates</td>
<td>Paul Söderlind, Ulf Söderström and Anders Vredin</td>
<td>2003:147</td>
</tr>
<tr>
<td>Inflation, Markups and Monetary Policy</td>
<td>Magnus Jonsson and Stefan Palmqvist</td>
<td>2003:148</td>
</tr>
<tr>
<td>Financial Cycles and Bankruptcies in the Nordic Countries</td>
<td>Jan Hansen</td>
<td>2003:149</td>
</tr>
<tr>
<td>Bayes Estimators of the Cointegration Space</td>
<td>Mattias Villani</td>
<td>2003:150</td>
</tr>
<tr>
<td>Business Survey Data: Do They Help in Forecasting the Macro Economy?</td>
<td>Jesper Hansson, Per Jansson and Mårten Löf</td>
<td>2003:151</td>
</tr>
<tr>
<td>The Equilibrium Rate of Unemployment and the Real Exchange Rate: An Unobserved Components System Approach</td>
<td>Hans Lindblad and Peter Sellin</td>
<td>2003:152</td>
</tr>
<tr>
<td>Bank Lending Policy, Credit Scoring and the Survival of Loans</td>
<td>Kasper Roszbach</td>
<td>2003:154</td>
</tr>
<tr>
<td>Internal Ratings Systems, Implied Credit Risk and the Consistency of Banks’ Risk Classification Policies</td>
<td>Tor Jacobson, Jesper Lindé and Kasper Roszbach</td>
<td>2003:155</td>
</tr>
<tr>
<td>Monetary Policy Analysis in a Small Open Economy using Bayesian Cointegrated Structural VARs</td>
<td>Mattias Villani and Anders Warne</td>
<td>2003:156</td>
</tr>
<tr>
<td>Intersectoral Wage Linkages in Sweden</td>
<td>Kent Friberg</td>
<td>2003:158</td>
</tr>
<tr>
<td>Do Higher Wages Cause Inflation?</td>
<td>Magnus Jonsson and Stefan Palmqvist</td>
<td>2004:159</td>
</tr>
<tr>
<td>Why Are Long Rates Sensitive to Monetary Policy</td>
<td>Tore Ellingsen and Ulf Söderström</td>
<td>2004:160</td>
</tr>
</tbody>
</table>