Growth and real business cycles in Vietnam and the ASEAN-5. Does the trend shock matter?

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Growth and real business cycles in Vietnam and the ASEAN-5.
Does the trend shock matter?∗

Binh T. Pham1, Hector Sala2, José I. Silva3

Abstract

We examine Vietnam’s economy together with its closest trade partners. We show that capital accumulation has been the primary growth engine since the start of its transition to the pro-market economy in 1986–the Doi Moi. We also show that the cyclical behavior of its macro aggregates is similar to the one of its ASEAN-5 peers and other developing countries. We extend the standard small-open-economy RBC model by considering habit persistence and government consumption which allows a close match of the moments of the growth variables. At the business cycle frequency, transitory productivity shocks account for approximately one-half of Vietnam’s output variance, while country-risk and non-transitory productivity shocks account to close to one-fifth each. Regarding Solow residual’s volatility, we find that the trend component merely accounts for 12% of this variance in Vietnam, while in Thailand it is only 6%. These findings refute “the cycle is the trend” hypothesis in Aguiar and Gopinath (2007), and align to those in García-Cicco, Pancrazi, and Uribe (2010) and Rhee (2017), in which the stationary component is overwhelmingly dominant. We claim that technological progress and productivity-enhancing measures are fundamental for Vietnam’s economy to sustain a high growth.

JEL Classification: E32, F43, 053

Keywords: Vietnam, ASEAN, DSGE, real business cycles, trend shock, growth

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1. Introduction

This paper examines Vietnam’s economy from a real business cycle (RBC) perspective, and compares its performance to that of ASEAN-5’s economies (Indonesia, Malaysia, the Philippines, Singapore, and Thailand). Vietnam provides a compelling case of analysis since it is a young and dynamic economy that has achieved many socio-economic successes over the past three decades.

The contribution of the paper is threefold. First, we extend the RBC models developed by Aguiar and Gopinath (2007) (AG) and García-Cicco, Pancrazi, and Uribe (2010) (GPU). We show that our extended setup provides a better account of the facts. In particular, we are able to match the downward slope autocorrelation between output and the trade balance (as percent of GDP), which was an unresolved issue in AG’s model. Second, we account for the sources of growth since the implementation of the Doi Moi in 1986, and show that the characteristics of Vietnam’s business cycles are not much different from its ASEAN-5 peers nor other emerging market economies. Third, we provide evidence that Vietnam’s business cycles have been mainly driven by transitory total factor productivity (TFP) shocks, rather than by trend innovations as claimed by AG for a set of 13 emerging economies. Given that the non-stationary component of the TFP shock only accounts for 12% of the Solow’s residual volatility (6% in Thailand), for these cases we refute their view that “the cycle is the trend” in emerging markets.

Our first task is to account for the supply-side factors that have driven Vietnam’s economic growth. We show that TFP accounts for one third of economic growth on average in 1981-2015, while it falls in the 2000s to less than one fifth. In addition, capital accumulation becomes the main driver of growth since 1992, with human capital rising its contribution. Then, we use well-known filtering techniques (e.g., Prescott 1986; King and Rebelo 1999; and Stock and Watson 1999, to name a few) to account for Vietnam’s business cycle fluctuations in the period from 1986 to 2015. We find that the business cycle characteristics of Vietnam’s national account components are essentially consistent with those reported by the literature for other emerging markets. Within this context, a significant difference is that Vietnam’s cyclical output fluctuation is less than half of the ASEAN-5’s average.

Provided with this information, we aim at quantifying the exogenous forces that have shaped the dynamics of Vietnam’s growth aggregates. Hence, we develop and estimate a dynamic stochastic general equilibrium (DSGE) RBC model to test the power of RBC theory in explaining Vietnam’s economic fluctuations.

In their highly influential work, AG asserted that the business cycle facts of a sample of thirteen developing countries can be adequately captured by a standard small open economy RBC model equipped with some real frictions. Specifically, they argued that the behaviors of consumption and the trade deficit depend on the nature of shocks to output growth. If a positive productivity shock is temporary, the resulting shift in consumption will lead to a proportional increase in output. In contrast, a permanent shock not only raises the current income but also gives rise to reduced savings (or investment), inducing a negative trade balance. AG showed that non-stationary shocks to productivity bear the main responsibility for output growth variations and suggested that in emerging markets the cycle is the trend.
Unsurprisingly, this view has received as many supports as objections in subsequent studies. For example, Suzuki (2018a,b) provided evidence that the business cycles of two emerging economies such as Serbia and South Africa are driven by the trend. Miyamoto and Nguyen (2017), employing a sample of seventeen small open economies, found that even if the trend role is not as important as in AG, the average contribution of trend productivity shocks to economic fluctuations is slightly above 30%.

On the other side, GPU pointed out that AG-type models could neither replicate the downward slope of the trade-balance-to-output autocorrelation function nor explain some crucial moments of the longtime series of Argentina and Mexico. GPU augmented AG’s model with financial frictions and country risk shocks and claimed that non-stationary productivity shocks only contributed by a small fraction to the output variance. Along the same line, Boz et al. (2011), Alvarez-Parra et al. (2013) and Rhee (2017), among others, provided support to GPU’s view. For instance, by considering a recursive utility function and an endogenous risk premium channel, Rhee (2017) showed that transitory productivity shocks significantly drive Korea’s economy.

Taking an intermediate viewpoint, Cao, L’Huillier, and Yoo (2016) postulate that two conditions for permanent shocks to dominate the cycle are an insensitive risk premium at any debt level, and time-separable preferences. Whereas the second condition can be satisfied by a suitable utility function, the absence of risk premium reaction causes the AG hypothesis to hold only in extremely unrealistic cases. In the same vein, Durdu (2013) concludes, after revising the literature, that the explanatory power of trend and/or interest rate shocks are magnified if the model contains a rich friction structure.

Accordingly, departing from the models by AG and GPU, we enrich GPU’s financial friction setting by incorporating into the period utility function: (1) internal habit persistence (Boldrin et al., 2001), and (2) government consumption (Christiano and Eichenbaum, 1992). In this way, our proposed RBC model outperforms GPU’s financial friction specification when reproducing the moments of the growth variables at the same time that matches the low value of trade-balance-to-output autocorrelation (0.18 after four lags). Although the presence of habit formation improves the moment matching capability, we acknowledge that the short-run observational dynamics cannot be emulated entirely. Nevertheless, neither the plain RBC nor the AG-type model is able to deliver a better performance.

The long-run variance decomposition reveals that transitory productivity shocks explain approximately 50% of Vietnam’s output volatility. Moreover, the transitory standard deviation is higher than its non-stationary counterpart by 25%. Another 41% of the variance is accounted by the exogenous risk premium and trend shocks, which have a similar contribution. While shocks to consumption preferences absorb the remaining 9% of the variance. These findings critically imply that AG’s claim that the cycle is the trend cannot be sustained in the case of Vietnam, where non-stationary shocks to productivity only account for 12% of the Solow residual’s volatility.

In order to gain insight into policy implications, we do counterfactual simulations under three scenarios. First, we simulate the trajectories of output growth and the trade-balance-to-output ratio by sequentially turning off
several exogenous processes: (i) by disabling trend (transitory) productivity shocks or interest rate shocks; (ii) by substituting the actual productivity shocks in 2008 – 2013 by the values they took in 1992 – 1997; and (iii) by assuming no interest rate shocks in 2008 – 2013 as if no (financial) crisis had taken place. The outcome of these counterfactual analyses is twofold. First, transitory productivity shocks have a significant impact on Vietnam’s income growth but not on trade balance. Neither trend nor productivity shocks can explain the huge trade deficits experienced in 2007 – 2010. Instead, interest rate shocks greatly govern the trade balance and help stabilize the growth path of Vietnamese output. This second result, which is striking since it unveils strong real effects of Vietnam’s proactive monetary policy in the past two decades, is consistent with the findings in Anwar and Nguyen (2018). In addition, our simulations strengthen Huynh et al. (2017)’s claim that the Vietnamese monetary policy was unable to counterbalance the economic downturn through the managing of interest rates, and that loan supply should have been directed toward productivity generating sectors.

We conclude the analysis by examining Thailand, which is the main competitor of Vietnam and has been the star economy within the ASEAN-5 historically. The explanatory power of both the transitory and trend shocks is relatively weak because each productivity innovation accounts only for one-fourth of the output growth variance. In this context, the trend component of the shock only accounts for 6% of Solow residual’s variance. In addition, in contrast to Vietnam, but similarly to Korea ((Rhee, 2017), country risk premium innovations appear to govern Thailand’s business cycles. We therefore argue that Thailand’s economy was more vulnerable to international externalities than Vietnam since the latter has stricter capital flow controls, and its economy is de-facto a non-free market economy.

The rest of the paper is structured as follows. Section 2 is devoted to the stylized facts of Vietnam and ASEAN-5 countries. Section 3 presents our RBC model. Sections 4 and 5 contain the estimation and counterfactual experiments for Vietnam. Section 6 briefly focuses on Thailand. Section 7 concludes.

2. Business cycle analysis

2.1. Data and Background

Macroeconomic research on Vietnam’s economy is challenging due to limited data availability. For output, the most recent quarterly data just covers years after 2000, but not all the other main aggregates are publicly provided. We collect annual aggregate data from the United Nations Statistical Division (UNSD) because the time coverage is long enough to identify the meaningful business cycles. To economize notation, we define a vector of main aggregate variables \( \mathbf{z} = (y, c, i, g, x, m, h, tby) \) corresponding, respectively, to (output, private consumption, investment, government consumption, exports, imports, and the trade-balance-to-output ratio); in turn, \( tby \) expresses the ratio of net exports over output. Table 1 summarizes all main aggregate growth rates for Vietnam and the average of the ASEAN-5 economies.

Vietnam’s real GDP per capita has steadily grown from a low of $200 US in the 1980s (constant 2005 USD) to six times higher over the past three decades (in nominal terms it reached more than $2100 US in 2015).
This marked the success of the Doi Moi program, an economic renovation strategy, which transformed Vietnam from one of the world’s poorest countries in the beginning of 1990s to a middle-income one in less than twenty-five years (World Bank, 2013, 2016).

Table 1 documents this success in comparison to the ASEAN-5 countries as Vietnam economy experienced higher growth rates in GDP as well as in all demand components with respect to the ASEAN-5 average. Taking as reference the whole period, 1986-2015, Vietnam’s real income growth sustained a higher level than the mean of the ASEAN-5 countries, 4.84% and 3.74% respectively. This positive differential was small prior to 2000, when the economy was still suffering from economic sanctions due to the Vietnam War consequences. However, the Bilateral Trade Agreement (BTA) between Vietnam and the US signed in 2001, and the subsequent Free Trade Agreements (FTA) with other East Asian countries (e.g. Korea, Japan, and China) in later years led to a flourishing economy. This explains the much larger differential in 2001-2015 (5.29% and 3.69%, respectively). In addition, Vietnam’s economy was much less volatile in 2001 – 2015 than in 1986 – 2000, as the standard deviation of GDP growth went down by a third.

<table>
<thead>
<tr>
<th>Country</th>
<th>VIETNAM</th>
<th>ASEAN-5</th>
</tr>
</thead>
<tbody>
<tr>
<td>Period</td>
<td>Stats</td>
<td>y</td>
</tr>
<tr>
<td>1986-2015</td>
<td>Mean</td>
<td>4.84</td>
</tr>
<tr>
<td></td>
<td>SD</td>
<td>1.74</td>
</tr>
<tr>
<td></td>
<td>Min</td>
<td>0.27</td>
</tr>
<tr>
<td></td>
<td>Max</td>
<td>7.40</td>
</tr>
<tr>
<td>1996-2000</td>
<td>Mean</td>
<td>4.40</td>
</tr>
<tr>
<td></td>
<td>SD</td>
<td>2.27</td>
</tr>
<tr>
<td></td>
<td>Min</td>
<td>0.27</td>
</tr>
<tr>
<td></td>
<td>Max</td>
<td>7.40</td>
</tr>
<tr>
<td>2001-2015</td>
<td>Mean</td>
<td>5.29</td>
</tr>
<tr>
<td></td>
<td>SD</td>
<td>0.82</td>
</tr>
<tr>
<td></td>
<td>Min</td>
<td>3.99</td>
</tr>
<tr>
<td></td>
<td>Max</td>
<td>6.36</td>
</tr>
</tbody>
</table>

Standard errors are not reported.

Table 1: Growth rate (in percentage, %) of main aggregate variables: output (y), private consumption (c), government consumption (g), investment (i), export (x), and import (m), except for trade-balance-to-output ratio (tby). ASEAN-5 countries: Indonesia, Malaysia, Philippines, Singapore and Thailand.

It is also interesting to compare the relative behavior in the two five-year recession periods, 1997-2001 and 2008-2012, related to the Asian crisis and the global financial crisis (GFC). In the former, Vietnam’s economy displayed an impressive growth rate of 5.06% per annum (0.18% the ASEAN-5 economies), while in 2008 – 2012 it became more vulnerable with a loss of 0.7 percentage points with respect to the average in 2001 – 2015 (note that this is the highest loss within the ASEAN community). Note that in the post Asian crisis years, Indonesia was the most stable economy together with Vietnam, while Singapore displayed the highest volatility.

The standard growth accounting framework (Solow, 1957) is often used to decompose output growth into parts due to input factors (i.e., capital and labor in the canonical Cobb-Douglas production function) and the Solow residuals. The latter component, the so-called total productivity factor (TFP), summarizes all
information about technological progress and other unexplained elements. Using the classical methodology in Bosworth and Collins (2003) and the Penn World Table (PWT) 9.0 datasets, in Table 2 we report the TFP growth rate and its volatility for Vietnam and the five ASEAN countries. The variability of the linearly detrended TFP behaves differently both across countries and time periods. Vietnam’s TFP varied in a considerably narrower band than its ASEAN-5 peers, as its standard deviations were the smallest, especially in the 2000s when they were very modest (0.6%). In the context of neoclassical economics, this explains why per capita GDP growth in Vietnam was much less fluctuating than its peers in the last two decades. Conversely, Singapore appears to be the most fluctuating economy. Note, finally, that Indonesia and Thailand experienced larger TFP growth rates positive than Vietnam in 2001–2015 (3.4% and 2.2% against 2.0%). Later on we briefly examine Thailand’s economy, while the unique case of Indonesia has been analyzed in Lee and Hong (2012).

<table>
<thead>
<tr>
<th>Country</th>
<th>Solow residuals’ growth rate, %</th>
<th>Standard deviation (σ_{sr}), %</th>
</tr>
</thead>
<tbody>
<tr>
<td>Indonesia</td>
<td>2.15 (0.53) 2.80 (0.77) 2.14 (1.53) 3.44 (0.19)</td>
<td>3.55 (1.27) 3.94 (1.63) 5.36 (2.11) 0.97 (0.17)</td>
</tr>
<tr>
<td>Malaysia</td>
<td>1.41 (0.55) 0.99 (0.58) 1.19 (1.00) 0.78 (0.60)</td>
<td>3.62 (0.69) 3.01 (0.74) 3.62 (1.15) 2.19 (0.58)</td>
</tr>
<tr>
<td>Philippines</td>
<td>1.32 (0.51) 0.39 (0.38) -0.29 (0.61) 1.07 (0.39)</td>
<td>3.41 (0.68) 1.99 (0.21) 2.29 (0.30) 1.41 (0.23)</td>
</tr>
<tr>
<td>Singapore</td>
<td>1.67 (0.53) 1.87 (0.74) 2.53 (0.97) 1.21 (1.12)</td>
<td>3.50 (0.42) 3.89 (0.52) 3.65 (0.86) 4.00 (0.72)</td>
</tr>
<tr>
<td>Thailand</td>
<td>1.37 (0.45) 1.77 (0.66) 1.35 (1.14) 2.17 (0.68)</td>
<td>2.96 (0.47) 3.35 (0.66) 3.98 (0.95) 2.40 (0.30)</td>
</tr>
<tr>
<td>Vietnam</td>
<td>1.96 (0.38) 1.90 (0.21) 1.85 (0.39) 1.97 (0.16)</td>
<td>2.54 (0.45) 1.10 (0.15) 1.42 (0.18) 0.59 (0.11)</td>
</tr>
</tbody>
</table>

Standard errors in parentheses are estimated by GMM estimator. Cyclical standard deviation obtained by regressing TFP over time trend $t$. That is $sr_t = sr_0 + \lambda t + \varepsilon_{sr}^t$; where $sr_t$ is the estimated Solow residuals from the Cobb-Douglas function, then $\sigma_{sr} = \sqrt{Var(\varepsilon_{sr}^t)}$.

Table 2: Growth rates and cyclical volatility of Solow residuals. Source: Penn World Table 9.0.

2.2. Business cycle facts

The growth accounting exercise so far provides the contribution of supply-side factors to output growth, but it cannot describe the variability of output in the short-run nor its co-movements with other aggregate variables. Those fluctuations have been defined as temporary deviations from some secular growth path over time (or the so-called “trend”), which should be a fitted smooth curve (Prescott, 1986). Accordingly, one can separate the transitory part out of the trend in several ways. In this paper we consider four standard econometric techniques which are first-differencing, linear regression, Hodrick and Prescott (1997) (HP) filter and the one-sided variant of HP filter (Stock and Watson, 1999). Note that elsewhere in the text the small case letter variables (e.g., $z_t$) denote natural logarithms of the respective level, which are expressed in capital letter ($Z_t$). Hence, the cyclical component should be interpreted as the percentage deviation from its trend.

Business cycle analysis involves computing the standard deviations of the cyclical main aggregate components ($\sigma_{j \in 3}$) and their relative ratios with respect to output ($\sigma_j / \sigma_y$). We apply all four different filters discussed above to extract cyclical signals out of the trends. The outcome is presented in Table 3.
At first glance, the HP filters provide the smoother paths, which are close to the demeaned first-differencing series and contrast with the oscillatory pattern resulting from the linear-quadratic filter. For Vietnam the latter delivers output standard deviations roughly 1.5 to 2.5 times larger than the HP and first-differencing counterparts (see Figure 1). According to the results from this estimation, Vietnam’s output oscillates around the trend by 4.75% per year on average so that two business cycles are clearly identified over our reference Doi Moi period. A first one in 1986 – 1997 and a second one in 1997 – 2008. Then, in the aftermath of the GFC, Vietnam’s output has evolved below the trend. Another relevant outcome is the different behavior of the ASEAN-5 countries, where 10-year cycles involving periods 1975 – 1986, 1987 – 1998, and 1998 – 2008 are identified. Lastly, Vietnam has lower volatility in GDP but higher relative volatility in demand components with respect to ASEAN-5 countries.

Given the estimated relative moments, Vietnam’s economic regularities seem to be consistent with the RBC literature for emerging countries (e.g., Uribe and Schmitt-Grohé, 2017). Investment and foreign trade activities are by far the most volatile components, coinciding with the growth accounting evidence reported
before. This could be explained by the Vietnam’s intense demand of imported goods over the past decades because of the needs of high-tech manufacturing equipment, by-products for fabricating and assembling industries, electronic devices, automobiles, and sizable investments in public infrastructure.

Although the specific case of linear filtering would not support this conclusion, the consumption of Vietnamese households seems to fluctuate more than output, in line with the higher variance of consumption to output reported in the literature on emerging economies. One of the great successes of the Doi Moi was the subsequent increase in the living standard of the Vietnamese people as “more than 40 million people escaped poverty over the course of two decades” World Bank (2016). This fact, which is connected to the low starting base of household consumption, explains at least partially the higher consumption-to-output standard deviation ratios in Vietnam with respect to the ASEAN-5’s economies.

Vietnam’s export and import growth rates have sustained a notable pace of 10% per year since 2001, twice that of the ASEAN-5 countries, leading the degree of trade openness to expand from 1.13 at the beginning of the 2000s to the height of 2.4 by the end of 2015. However, Vietnam’s trade balance was negative over the whole thirty-year period, revealing weak competitiveness, over-consumption of imported goods, and vulnerability of the economy to adverse shocks, especially during the GFC in 2008 – 2012. Besides, the autocorrelation function of the trade-balance-to-output ratio presents a monotonically downward trend approaching 0.18 beyond the fourth order (see Table 4). This phenomenon is commonly observed in emerging markets but not all studies have succeeded in matching or reproducing it.

Looking at the results from the one-sided HP filter perspective, Vietnam’s exports and imports seem to be experiencing procyclical movements, as their first lagged and contemporaneous correlations with output are at moderate levels, 0.57 and 0.65 respectively. These figures are close to the ASEAN-5’s average as displayed in Table 4. Note that all autocorrelation coefficients almost die out after two years, yet they have three times more volatility than output. This is indicative of the fact that Vietnam and ASEAN-5’s foreign trades were fairly sensitive to international economic conditions.

Household consumption and investment present the expected procyclical behavior. Their contemporaneous correlations with output are in the range of [0.5, 0.6], and investment interestingly exhibits a “time-to-build” effect as its first and second-order correlations are, respectively, as high as 0.74 and 0.77. Correspondingly, lead relationships between investment and output are poor, and the second-order lead seems to be uncorrelated. On the contrary, ASEAN-5’s private consumption and investment are procyclical and moderately persistent, as their contemporaneous correlations and first-order auto-correlations are roughly 0.7.

In contrast, both Vietnam and ASEAN-5’s public demand for goods and services behave acyclically and display some persistence, as indicated by their first-order auto-correlations and contemporaneous correlation with output which are, respectively, above 0.70 and below 0.28 (but positive). The same reading applies to the trade-balance-to-output ratio variable.

To conclude, the above set of business cycle facts show that the characteristics of Vietnam’s economy are similar to those of the ASEAN-5 countries and reconcilable with the findings of the RBC literature on other
The production function is defined as:

\[
Y_t = A_t K_{t-1}^\alpha (X_t h_t)^{1-\alpha}
\]  

(1)

3. Model

The natural question arising from the previous analysis refers to the kind of technological shocks that are most relevant in driving the economic fluctuations of small open emerging economies like Vietnam and the ASEAN-5 peers. To respond to this question, we develop a small open economy DSGE-RBC model aiming to quantify the exogenous forces that shape the dynamics of Vietnamese growth aggregates (and Thailand’s ones in a final comparative exercise).

3.1. Setting the economic environment

The model is an extension of GPU’s financial friction specification which adds internal habit persistence (Boldrin, Christiano, and Fisher, 2001) and government consumption into the period utility function (Christiano and Eichenbaum, 1992). In what follows, we opt for the end-of-period notation since it is naturally compatible with Dynare’s coding convention.

We assume a symmetric, single good economy endowed with a constant-returns-to-scale Cobb-Douglas production technology. The production function is defined as:

\[
Y_t = A_t K_{t-1}^\alpha (X_t h_t)^{1-\alpha}
\]  

(1)
where $X_t$ represents labor-augmenting technological change which has a cumulative effect as noted in AG. Thus, $X_t = g_t X_{t-1}$; where $g_t$ is the productivity’s gross rate of growth so that $\log(g_t/\mu_g) = \rho_g \log(g_{t-1}/\mu_g) + \xi_t^g$, with $\mu_g$ being the gross long-run growth rate and $|\rho_g| < 1$. The single trend shock $\xi_t^g$ is assumed normally distributed with variance $\sigma_g^2$, $\xi_t^g \sim N(0, \sigma_g^2)$.

Variable $a_t \equiv \log(A_t)$ denotes the transitory productivity process following the usual AR(1) propagation mechanism such that $a_t = \rho_a a_{t-1} + \varepsilon_t^a$, with $|\rho_a| < 1$ and $\varepsilon_t^a \sim N(0, \sigma_a^2)$.

Capital stock accumulation is subject to the following law of motion:

$$K_t = (1 - \delta)K_{t-1} + h_t - \frac{1}{2} \left( \frac{K_t}{K_{t-1}} - \mu_g \right)^2 K_{t-1}$$

(2)

where $\delta$ represents the rate of depreciation and $\phi$ is the parameter to be estimated. Note that the last term on the right hand side governs the capital adjustment costs.

The instantaneous utility function takes the Greenwood, Hercowitz, and Huffman (1988) (hereafter GHH) form as in GPU because it is well-known that GHH preferences generate the excess volatility of consumption over output and counter-cyclical net exports (see Correia et al., 1995). Thus,

$$U(C_t^*, h_t) = \left[ \frac{C_t^* - \theta \omega^{-1} X_{t-1} h_t^{1/3}}{1 - \eta} - 1 \right]^{1-\eta}$$

(3)

where $C_t^* = v_t C_{p,t} - \gamma C_{p,t-1}^* + \pi C_{g,t}^*$, with $C_p$ and $C_g$ being private and public consumption, respectively; $\eta > 0$ and $\eta \neq 1$; and $v_t$ denotes an exogenous and stochastic preference shock. The existence of $C_{p,t-1}$ and $\gamma$ in the utility function accounts for internal habit persistence, implying that household consumption has a time-non-separable structure as past decisions affect the present utility. Two parameters, $\theta > 0$ and $\omega > 0$, control the disutility of work and need to be calibrated to the normalized long-run hours-work of the target country. The involvement of government consumption in the instantaneous utility (3) is determined by $\pi \in [0, 1]$, which represents the elasticity of substitution between private and public consumption. We model $C_{g,t}^* = \zeta_{cg} \xi_{cg} Y_t$, with the spending shock $\xi_{cg}$ and a constant ratio of $\zeta_{cg}$.

Specifying the utility function as in (3) has several non-trivial implications. First, habit formation generally improves the explanatory power of the DSGE models by allowing consumption smoothness (see e.g., Fuhrer 2000; Christiano et al. 2005; and Ravn et al. 2006, to name a few). Second, the present internal setting for habit persistence is advocated by Constantinides (1990) and Boldrin et al. (2001) as their studies show that “internal habit” is better than the “keeping-up-with-the-Joneses” counterpart in jointly explaining the risk premium puzzle and business fluctuations that small open economies likely encounter. Lastly, GHH preferences give rise to the complementarity between consumption and hours. In our case, by incorporating government spending into the period utility we allow for partial ($0 < \pi < 1$) or complete ($\pi = 1$) complementarity between private and public spending. Note that this is in contrast to GPU’s financial friction model, which implicitly imposes $\pi = 0$. 

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The two AR(1) processes, \( \nu_t \) and \( \xi_{cg,t} \), perturb the present household and government consumption as follows:

\[
\begin{align*}
\log(\nu_t) &= \rho_\nu \log(\nu_{t-1}) + \varepsilon_{\nu,t} \\
\log(\xi_{cg,t}) &= \rho_{cg} \log(\xi_{cg,t-1}) + \varepsilon_{cg,t}
\end{align*}
\]

where \( \varepsilon_{\nu,t} \sim N(0, \sigma^2_\nu) \) and \( \varepsilon_{cg,t} \sim N(0, \sigma^2_{cg}) \).

The household budget is:

\[
\frac{B_t}{1 + r_t} + Y_t = C_{p,t} + C_{g,t} + I_t + B_{t-1}
\]

(4)

where \( B_t \) and \( r_t \) are the stock of debt and domestic interest rate at time \( t \), respectively. The trade-balance-to-output ratio in the model is defined as \( TB_t = (B_{t-1} - B_t/(1 + r_t))/Y_t \), so that the negative value of \( B_t \) represents the economy-wide indebtedness (i.e., investment over saving). The market clearing condition is written as:

\[
(Y_t - C_{p,t} - C_{g,t}) - I_t = TB_t
\]

(5)

Regarding the bond discount rate \( r_t \), Rhee (2017) considers an endogenous risk premium channel—proposed by Neumeyer and Perri (2005)—as an alternative to GPU’s configuration. The latter approach postulates a domestic interest rate that is the sum of the world interest rate \( r^* > 0 \) (assumed to be constant), the country’s risk premium, and an exogenous shock to the country’s premium \( \mu_t \). Since we are using the GPU’s model as benchmark, the rule for the exogenous risk premium channel is restated as:

\[
r_t = r^* + \psi(e^{B_t/X_t - \bar{b}} - 1) + e^{(\mu - 1)} - 1
\]

(6)

where \( \log(\mu_t) = \rho_\mu \log(\mu_{t-1}) + \varepsilon_{\mu,t} \), \(-1 < \rho_\mu < 1\) with \( \varepsilon_{\mu,t} \sim N(0, \sigma^2_\mu) \); \( \psi \) is a parameter governing the debt elasticity; and \( \bar{b} \) is the steady-state level of governmental outstanding debt.

Our model is richer than GPU’s financial friction setup in the number of exogenously stochastic processes as it allows to distinguish shocks to the pure-time discount rate \( (\xi_p) \) from shocks to the current consumption \( (\nu_t) \). The former is also known as a preference-shifter since it lets parameter \( \beta \) vary across time, while the latter is the same as in GPU. The representative household thus seeks to maximize (7) subject to constraints (2) and (4):

\[
E_0 \sum_{t=0}^{\infty} \xi_{\mu,t} \beta^t U(C_t, h_t)
\]

(7)

11
As a consequence, the Lagrangian function is:

$$E_0 \sum_{t=0}^{\infty} \beta^t \xi_{p,t} \left\{ \frac{v_tC_{p,t} - \gamma C_{p,t-1} + \pi \omega^{-1} X_{t-1} h_t^{\omega}}{1 - \eta} - \Lambda_t \right\}$$

with Lagrangian multiplier $$\Lambda_t = \lambda_t X_{t-1}^{-\eta}$$.

Along with equations (1), (2), (4) and (6); and the six AR(1) exogenous stochastic shocks, the first-order conditions of (8) give us the set of equilibrium conditions (9) – (12) for:

### Consumption

$$\xi_{p,t} \frac{v_t}{(v_tC_{p,t} - \gamma C_{p,t-1} + \pi C_{g,t} - \theta \omega^{-1} h_t^{\omega})^\eta} - \frac{\lambda_t}{X_{t-1}^\eta} = \frac{\gamma \beta \xi_{p,t+1}}{(v_{t+1} C_{p,t+1} - \gamma C_{p,t+1} + \pi C_{g,t+1} - X_t \theta \omega^{-1} h_{t+1}^{\omega})^\eta}$$

### Hours-work

$$\frac{\lambda_t (1 - \alpha) a_t (X_t h_t)^{1-\alpha} K_t^{\alpha}}{X_{t-1}^\eta h_t} = \frac{\theta X_{t-1} v_t h_t^{\omega-1}}{(v_tC_{p,t} - \gamma C_{p,t-1} + \pi C_{g,t} - X_{t-1} \theta \omega^{-1} h_t^{\omega})^\eta}$$

### Capital

$$\frac{\lambda_t \xi_{p,t}}{X_{t-1}^\eta} \left[ \left( \frac{K_t}{K_{t-1}} - \mu_e \right) \phi + 1 \right] = \frac{\lambda_{t+1} \xi_{p,t+1}}{X_t^{\eta}} \beta \left[ \alpha a_t \left( \frac{X_t h_t^1}{K_t} \right)^{1-\alpha} - \frac{\phi}{2} \left( \frac{K_{t+1}}{K_t} - \mu_e \right)^2 + \frac{\phi}{2} \left( \frac{K_{t+1}}{K_t} - \mu_e \right) + 1 - \delta_0 \right]$$

### Domestic bond

$$1 + r_t = \frac{\lambda_t}{\lambda_{t+1}} \frac{\xi_{p,t}}{\xi_{p,t+1}} \frac{g_t^\eta}{\beta}$$

To see the difference between internal and external consumption persistence, replace $$C_{p,t-1} \equiv C_{p,t-1}$$ (the aggregate private consumption at time $$t - 1$$) in the period utility function; as a consequence, the optimizing household decides her current consumption taking the previous aggregate private spending as given so that (9) reduces to:
\[
\frac{\lambda_t}{X_t^\eta} = \frac{\nu_t}{(\nu_t \mathcal{C}_{p,t} - \gamma \mathcal{C}_{p,t-1} + \pi \mathcal{C}_{g,t} - X_{t-1} \theta \omega^{-1} h^\omega) \eta} \tag{13}
\]

It is obvious that the preference shifter \(\xi_p\) does not perturb Lagrangian multiplier \(\lambda_t\) in (13). This implies that an external habit specification is inefficient in our setting.\(^4\)

The model above is non-stationary and will not converge to the balance-growth path because the output \(Y_t\) increases over time by the factor \(X_t\) in equation (1). It is thus necessary to detrend all equilibrium conditions, which we do following AG.\(^5\)

### 3.2. The long-run equilibrium

We characterize the long-term relationships among (detrended) national income identity variables by ruling out the presence of all exogenous shocks. Given that the domestic interest rate and the world interest rate are identical in equilibrium, we have:

\[1 + r^* = \frac{\mu_g^\eta}{\beta} = 1 + r\]  \tag{14}

and

\[
\frac{h}{k} = \frac{1}{\mu_g^\eta} \left[ \frac{\mu_g^\eta / \beta + \delta - 1}{\alpha} \right]^{1/(1 - \alpha)} = \frac{1}{\mu_g^\eta} \left[ \frac{r + \delta}{\alpha} \right]^{1/(1 - \alpha)} \tag{15}
\]

Equation (15) expresses the labor-to-capital relationship. Since \(0 < \alpha < 1\), the ratio \(h/k\) would increase with \(r\), ceteris paribus. There are three ways to lift up the domestic interest rate \(r\) in equilibrium (all else being equal). First, by calibrating under a higher the value of \(\eta\); secondly, under a higher long-run labor-augmenting growth rate \(\mu_g\); third, under a lower subjective discount factor \(\beta\).

The feature that distinguishes our approach from AG-type models is the presence of habit formation (\(\gamma\)), which changes the way hours-work \(h_t\) behaves:

\[
h = \frac{\theta}{\mu_g (1 - \alpha) \left( 1 - \frac{\nu_t}{\mu_f} \right)} \left[ \frac{r + \delta}{\alpha} \right]^{\alpha/(1 - \alpha)} \tag{16}
\]

We calibrate \(\omega\) to a value greater than unity, which is the standard in RBC literature. Subsequently, an increase in the internal habit formation coefficient, \(\gamma\), decreases the steady-state labor-supply to a lower value.

\(^4\)Identification test (Iskrev, 2010) reports \(\xi_{p,t}\) and \(\nu_t\) are pairwise multi-collinearity if \(\gamma\) is external.

\(^5\)We follow the standard convention according to which a stationary variable – represented by a lowercase letter – is equivalent to the ratio of that variable (dividend) with respect to \(X_{t-1}\) (divisor). Following the end-of-period convention, the variable determined at time \(t\) will be adjusted by the factor \(X\) at time \(t - 1\), and so on.
underpinning the household resistance to unanticipated changes. Likewise, the lower value of the subjective
discount factor $\beta$ will decrease the hours-work, since the household may become impatient. Accordingly, the
ratios $(k/y)$, $(i/y)$, and $(c_p/y)$, which are functions of $(h/k)$ as defined in equations (17) – (19), are directly
affected by the deep parameters implied in equations (15) and (16),

\[
\frac{k}{y} = \mu_k^{1-2\alpha} \left( \frac{h}{k} \right)^{1-\alpha} \\
\frac{i}{y} = \frac{k}{y} \left( 1 - \frac{1-\delta}{\mu_g} \right) \\
\frac{c_p}{y} = 1 - \zeta_{cg} - \zeta_b \left( \frac{1}{\mu_g} - \frac{1}{1+r} \right) - \frac{k}{y} \left( 1 - \frac{1-\delta}{\mu_g} \right) \\
\frac{x-m}{y} = \frac{tb}{y} = \zeta_b \left( \frac{1}{\mu_g} - \frac{1}{1+r} \right)
\]

where $\zeta_{cg}$ and $\zeta_b$ are, respectively, the shares of government consumption and the stock of bonds in the
steady-state.

The model is the collection of equilibrium equations (9) – (12), six AR(1) processes, and equations (1), (2),
(4) and (5). It is then solved by the second-order solution algorithm of Schmitt-Grohé and Uribe (2004)
before proceeding estimation.

4. Estimation and discussions

We estimate the model using the standard Bayesian Markov Chain Monte Carlo (MCMC) estimator which,
after Smets and Wouters (2003), is used in most recent DSGE literature and is available in Dynare and similar
packages. The Bayesian estimation lends itself to the Bayes rule for the conditional distribution of a set of
estimating parameters $\theta \in \Theta$ given observational data $Y$. The core formula is:

\[
\Gamma(\theta|Y) = \frac{f(Y|\theta)\Gamma(\theta)}{f(Y)}
\]

where $\Gamma(\theta|Y)$ is the so-called posterior probability distribution function of the parameter $\theta$ (or posterior
distribution shortly) conditional on observational data $Y$. The prior distribution $\Gamma(\theta)$ is the unconditional
probability distribution of $\theta$, whilst the likelihood function $f(Y|\theta)$ is defined as in classical econometric
methods. The last component, $f(Y)$, is the marginal likelihood defined as: $f(Y) = \int f(Y|\theta)\Gamma(\theta)d\theta$.

Methodologically, a Bayesian estimation requires three key elements: advanced knowledge of prior specifica-
tions, a suitable filter for likelihood evaluation at every observational data point, and the Metropolis-Hastings
Monte Carlo (MHMC) simulation algorithm (see also An and Schorfheide 2007; Fernández-Villaverde 2010;
and Herbst and Schorfheide 2015 for in-depth technical expositions).
4.1. Configuration

A time unit is meant to represent a year in the model. Econometrically, not all model “deep” parameters are estimated, as some of them will be calibrated to the commonly used values within the RBC literature (see e.g., Cooley and Prescott 1995; King and Rebelo 1999; and Schmitt-Grohé and Uribe 2003, among many others). For instance, the depreciation rate and labor-supply elasticity are micro-based parameters which are not intrinsically supported by the main aggregate information.

Table 5 reports the choices of a number of calibrated parameters based on GPU and Vietnam long-run data. First, we set the value of $\omega$ to 2.0, which results in $\theta = 1.85$ so that the normalized value of hours-work is approximately one-fourth of a unit time-endowment. For consistency with earlier growth accounting evidence, we set the capital income share to $\alpha = 0.35$. The depreciation rate, $\delta$, and the curvature of the period utility function, $\eta$, take the common values of 10% and 2, respectively. Next, the ratio of government consumption, $\zeta_{cg}$, and stock of domestic bond to output, $\zeta_b$, are calibrated to 0.07 and $-0.36$, respectively. The former value is simply the sample average, but the latter is determined from the long-run trade-balance-to-output ratio of $-2.6\%$ in years before 2000.

We neither calibrate the subjective discount factor $\beta$ nor the long-term growth rate $\mu_g$ as Rhee (2017) and GPU do in their studies. Given the absence of evidence on Vietnam’s business cycles, they need to be estimated. To check for robustness and verify the result’s sensitivity to these estimates, we consider a range of values such that $\mu_g \in [0.03, 0.05]$ and $\beta^* \in [0.03, 0.07]$. The presence of government consumption in the period utility causes government consumption to reduce the total household utility whenever $0 \leq \pi < 1$. In turn, fixing $\pi = 1$ entails full complementarity between private and public consumption.\(^6\)

<table>
<thead>
<tr>
<th>Calibrated params</th>
<th>$\beta^*$</th>
<th>$\pi$</th>
<th>$\theta$</th>
<th>$\omega$</th>
<th>$\eta$</th>
<th>$\alpha$</th>
<th>$\delta$</th>
<th>$\mu_g$</th>
<th>$\zeta_{cg}$</th>
<th>$\zeta_b$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Value</td>
<td>0.03-0.07</td>
<td>1</td>
<td>1.85</td>
<td>2.00</td>
<td>2.00</td>
<td>0.35</td>
<td>0.10</td>
<td>1.048</td>
<td>0.07</td>
<td>-0.36</td>
</tr>
</tbody>
</table>

Table 5: Calibrated parameters

There are a total of sixteen structural parameters to be estimated, comprising six AR(1) coefficients, their corresponding exogenous stochastic disturbance variances, and $\beta, \psi, \gamma, \phi$. GPU suggested the sole use of uniform distributions to estimate the structural parameter space. Rhee (2017), however, estimates the AG-type model using a mixture of Beta, Gamma, and Inverse – Gamma distribution priors. As shown in Table 6 (first four columns), we adopt Rhee (2017)’s prior specifications.

It is worth emphasizing that an advantage of directly modeling non-stationary variables in the AG- and GPU-type models is that it allows for model-based detrending instead of an arbitrary selection amongst

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\(^6\)Even though it passes the Iskrev (2010)’s test, the identification strength of $\pi$ is weak (with a hardly distinguishable posterior from the prior distribution). Figure A4 and A5 provides a sensitivity analysis to study how the variability of parameter $\pi$ affects the modeled variables. The habit persistence coefficient appears to be the strongest parameter, while the elasticity of public spending $\pi$ only has considerable effects on the responses of $g_t$ and $\varepsilon_g$ with respect to two shocks $\varepsilon_{cg}$ and $\varepsilon_p$. Hence, contrary to what could be expected, the presence of public spending in the utility function does not have a relevant impact. This is in line with the conclusions in Cantore, Levine, and Melina (2014).
abundant filtering techniques used in data transformations for estimating deep parameters. Canova (2014) stresses on the consistency between model and data when both permanent and transitory shocks coexist. Along the same line, Canova and Ferroni (2011) and Ferroni (2011) show that structural estimates could be biased or distorted due to the wrong choice of time series filter or trend misspecification.

Our model has more shocks (six) than observed data (five), namely four growth rate aggregates and the trade-balance-to-output series. The direct links among data and model variables, i.e., the measurement equations, are defined as:

\[
\begin{align*}
    g_y, t & = \Delta \log(Y_t) = \log(y_t) - \log(y_{t-1}) + g_{t-1} - 1 \\
    g_{cp}, t & = \Delta \log(C_{p,t}) = \log(c_{p,t}) - \log(c_{p,t-1}) + g_{t-1} - 1 \\
    g_{cg}, t & = \Delta \log(C_{g,t}) = \log(c_{g,t}) - \log(c_{g,t-1}) + g_{t-1} - 1 \\
    g_{inv}, t & = \Delta \log(I_t) = \log(i_t) - \log(i_{t-1}) + g_{t-1} - 1 \\
    tby, t & = \frac{TB_t}{Y_t} = \frac{y_t - c_{p,t} - c_{g,t} - i_t}{y_t}
\end{align*}
\]

where the vector of observational data \((g_y, g_{cp}, g_{cg}, g_{inv})\) is the per capita annual growth-rate of \((output, private consumption, government consumption, investment)\); and \(tby\) represents trade-balance-to-output ratio. These linkages arise naturally from the model implying the model-based differencing data transformation. Following GPU, a measurement error is added to each observed variable to resolve filtering errors, data quality or even occasional model misspecification (Del Negro and Schorfheide, 2009). Besides, although the sample size could be taken to be a matter (we exercise Bayesian estimation over 1981 – 2015), it has been shown that relatively small sample sizes can produce valid Bayesian inference within the DSGE context (Fernández-Villaverde and Rubio-Ramírez 2004; and Benchimol and Fourçans 2017).

4.2. Results

4.2.1. Estimated parameters

All estimated parameters are reported in columns 5 - 9 of Table 6. The results indicate that the subjective discount rate is fairly moderate, of approximately 6.4% per annum (equivalent to 1.6% per quarter), implying that the stationary value of real interest rate \(r = \mu^\eta / \beta\) would be in the range of \([12.2\%, 16.2\%]\) corresponding to \(\eta = [1.2, 2.0]\) and \(\mu_g = 1.045\). As a sensitivity check, we run Bayesian estimations at each \(\eta = (1.25, 1.50, 1.75, 2.00)\), which still deliver the consistent interval of \(\beta^\ast \in [6.3\%, 6.9\%]\). The low value of \(\beta\) for Vietnam (0.94), when compared to the one obtained for the US by King and Rebelo (1999), \(\beta = 0.98\), indicates that a Vietnamese household is qualitatively more impatient than one living in an advanced country. This is not unreasonable in view of the severe difficulties regarding high inflation experienced by Vietnam in the 1980s and 1990s, and even in the aftermath of the GFC, regarding high inflation. Unlike most small open emerging economies, however, Vietnam is on its way to transforming from a closed and centralized economy to a pro-business and pro-market one.

<table>
<thead>
<tr>
<th>Estimated params</th>
<th>Prior</th>
<th>Posterior</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Dist.</td>
<td>Mean</td>
</tr>
<tr>
<td>$\mu_g$</td>
<td>Beta</td>
<td>1.05</td>
</tr>
<tr>
<td>$\beta^* = \frac{1}{\hat{p}} - 1$</td>
<td>Beta</td>
<td>0.04</td>
</tr>
<tr>
<td>$\psi$</td>
<td>Gamma</td>
<td>0.50</td>
</tr>
<tr>
<td>$\gamma$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\phi$</td>
<td>Gamma</td>
<td>5.00</td>
</tr>
<tr>
<td>$\rho_a$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\rho_s$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\rho_p$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\rho_{cg}$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\rho_{\mu}$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\rho_v$</td>
<td>Beta</td>
<td>0.50</td>
</tr>
<tr>
<td>$\sigma_a$</td>
<td>IGamma</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma_s$</td>
<td>IGamma</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma_p$</td>
<td>IGamma</td>
<td>0.05</td>
</tr>
<tr>
<td>$\sigma_{cg}$</td>
<td>IGamma</td>
<td>0.05</td>
</tr>
<tr>
<td>$\sigma_{\mu}$</td>
<td>IGamma</td>
<td>0.05</td>
</tr>
<tr>
<td>$\sigma_v$</td>
<td>IGamma</td>
<td>0.05</td>
</tr>
<tr>
<td>$\sigma_{sy}$</td>
<td>IGamma</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma_{cp}$</td>
<td>IGamma</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma_{cp}$</td>
<td>IGamma</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma_{g}_{inv}$</td>
<td>IGamma</td>
<td>0.10</td>
</tr>
<tr>
<td>$\sigma_{by}$</td>
<td>IGamma</td>
<td>0.05</td>
</tr>
</tbody>
</table>

*Beta* denotes Beta distribution; (I)Gamma represents (inversed) gamma distribution. p-val is p-value (15% taper) of Geweke (1992) Convergence Test (see appendix for additional trace plots). The estimation is based on 1000,000 draws from the MHMC algorithm.
The estimated value of $\psi$ is 0.29 (much lower than GPU’s estimate for Argentinian data), reflecting that the Vietnamese domestic interest rate is quite insensitive to the debt level. In turn, capital adjustment costs $\phi \sim 4.0$ are remarkably close to those found in Latin-America by GPU, suggesting that fixed-capital formation in developing countries is, in general, a costly process to sustain growth targets.

Regarding the AR(1) processes, the six estimated coefficients $\rho_{(\cdot)}$ are within the plausible range of $[0.58, 0.78]$, in accordance with the use of annual data. The fluctuations of both stationary and non-stationary technological shocks are fairly small, 0.68% and 0.55%, respectively; whereas the variations of the temporal preference and consumption taste are comparatively large, 4.4% and 3.9%, respectively. These estimates are able to account for the excess volatility of Vietnamese household consumption compared to output.

Table 7: Vietnam: growth variable moments from Bayesian estimation, 1981 – 2015

<table>
<thead>
<tr>
<th>Moment</th>
<th>Actual</th>
<th>W. habit</th>
<th>Model</th>
<th>GPU</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\sigma_{gy}$</td>
<td>1.64</td>
<td>(0.21)</td>
<td>1.69</td>
<td>2.23</td>
</tr>
<tr>
<td>$\sigma_{gpc}/\sigma_{gy}$</td>
<td>1.35</td>
<td>(0.17)</td>
<td>1.21</td>
<td>1.84</td>
</tr>
<tr>
<td>$\sigma_{gc}/\sigma_{gy}$</td>
<td>2.03</td>
<td>(0.46)</td>
<td>2.24</td>
<td>2.58</td>
</tr>
<tr>
<td>$\sigma_{gct}/\sigma_{gy}$</td>
<td>6.84</td>
<td>(1.44)</td>
<td>5.31</td>
<td>6.24</td>
</tr>
<tr>
<td>$\sigma_{gy}$</td>
<td>2.61</td>
<td>(0.59)</td>
<td>2.82</td>
<td>3.91</td>
</tr>
</tbody>
</table>

Note: standard error in parentheses. AR$_1$ denotes first-order autocorrelation. W. habit is model with internal habit persistence; W/o is without habit coefficient.

Table 8: Simulated (auto) correlations of level variables.
4.2.2. Simulated results

The top block of Table 7 reports the striking performance of the model (column W. habit), as it can effectively reproduce the output variability and other important relative standard deviations. In the absence of internal habit formation (W/o habit), the predicted variance of output is higher than the actual one by 25%; the ratios \( \sigma_{gyp} / \sigma_{gy}, \sigma_{gim} / \sigma_{gy}, \sigma_{gim} / \sigma_{gy}, \sigma_{tby} / \sigma_{gy} \) are higher than those of the full model; and neither the downward slope of the trade-balance-to-output autocorrelation function (Figure 2) nor the excess variation of household consumption to output can be reproduced in the GPU’s basic setting. When GPU’s financial friction (FFR) is added, the output growth variance is over predicted while the \( \sigma_{ginv} / \sigma_{gy} \) is underestimated. Besides, the basic model has a tendency of generating strong procyclical growth rates of consumption and investment.

Failure to model consumption memory would worsen the model’s moment matching and the short-run dynamic behaviors. The estimated habit persistence \( \gamma \) is close to the mean value reported in the meta-analysis of (Havranek et al., 2017, Table 1), 0.27 and 0.30 respectively. This indicates that the habit persistence strength of Vietnamese households is as strong as those living in advanced countries. This strength underlies the gradual responses of Vietnamese private consumption and inflation to all policy shocks (Fuhrer, 2000).

We need to acknowledge that neither our proposed model nor GPU’s one can wholly recreate the short-run dynamics of Vietnam’s growth observables, as displayed in the two bottom blocks of Table 7. This can be explained by the relatively small sample timeframe we used in comparisons to the 100-year datasets of Mexico and Argentina in GPU’s study. On the other hand, Table 8 demonstrates that the model is sound in capturing the contemporaneous and first-order autocorrelations of the four aggregate levels, namely \( y, c_p, c_g \) and \( i \).

Finally, we highlight the failure of GPU-type models in replicating Vietnam’s cyclical moments due to: (i) the time-invariant subjective discount factor \( \beta \), which implies that optimizing agents do not adjust their forward-looking expectations over time; (ii) the absence of a mechanism that permitting inter-temporal consumption smoothing; and crucially (iii) the co-existence of habit in consumption and capital adjustment costs, which markedly reduces the volatility in investments and output (Khorunzhina, 2015): our estimation is eloquent in reporting increasing \( \sigma_{gy} \) and other relative ratios when optimizing agents have no consumption memory.

---

Havranek et al. (2017) explore 81 studies covering Australia, New Zealand, G-7 group, and many EU countries.
4.3. Variance decomposition

Using Kalman’s filter, we compute the long-term (unconditional) variance decomposition of the variables due to orthogonal shocks and compare them across models (Table 9). The results indicate that the contribution of a transitory TFP shock to the output growth variance amounts to 50% if we use our extended model (89% when using GPU’s one). Accordingly, it seems safe to conclude that Vietnam’s business cycles in 1981 – 2015 were not driven by trend innovations. This is in clear contrast to the results of AG for their set of 13 emerging market economies.

Shocks to trend productivity and the country risk premium account for about 20% each of the remaining output variability. This finding uncovers a novel relevant impact of risk premium shocks on economic growth. Moreover, the results suggest that the long-term fluctuations of investments and the trade-balance-to-output ratio are mainly driven by the country risk premium. Surprisingly, GPU’s financial friction contradicts this key finding by predicting a critical impact of the consumption preference shock on the trade balance. To illustrate the significant impact of risk premium $\epsilon_\mu$ on the observable $tby$, we decompose its variance over time, as displayed in Figure 3. This figure shows that most of the variation in the trade-balance-to-output ratio in 1999 – 2015 was due to the exogenous interest rate disturbances, $\epsilon_\mu$. 

Figure 2: Trade-balance-to-output ratio ($tby$) autocorrelation of Vietnam (left panel) and Thailand (right panel). VNM (THA) model refers to the our RBC model; VNM basic is the standard RBC prototype in GPU.
<table>
<thead>
<tr>
<th>MODEL</th>
<th>Output growth</th>
<th>Cons. growth</th>
<th>Gov. spending growth</th>
<th>Investment growth</th>
<th>Trade balance to GDP ratio</th>
</tr>
</thead>
<tbody>
<tr>
<td>Nonstationary tech</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>w. habit</td>
<td>18.67</td>
<td>7.90</td>
<td>8.65</td>
<td>8.53</td>
<td>5.84</td>
</tr>
<tr>
<td>w/o habit</td>
<td>26.60</td>
<td>7.00</td>
<td>3.97</td>
<td>11.77</td>
<td>9.10</td>
</tr>
<tr>
<td>GPU-FFR</td>
<td>[3.81]</td>
<td>[1.20]</td>
<td>[1.35]</td>
<td>[1.15]</td>
<td></td>
</tr>
<tr>
<td>Stationary tech</td>
<td>49.72</td>
<td>10.64</td>
<td>22.69</td>
<td>2.46</td>
<td>1.62</td>
</tr>
<tr>
<td></td>
<td>51.25</td>
<td>7.12</td>
<td>7.49</td>
<td>1.03</td>
<td>1.70</td>
</tr>
<tr>
<td></td>
<td>[88.71]</td>
<td>[10.18]</td>
<td>[3.63]</td>
<td>[5.16]</td>
<td></td>
</tr>
<tr>
<td>Preference shift</td>
<td>0.07</td>
<td>4.88</td>
<td>0.03</td>
<td>0.16</td>
<td>0.63</td>
</tr>
<tr>
<td></td>
<td>0.13</td>
<td>12.39</td>
<td>0.02</td>
<td>0.21</td>
<td>2.94</td>
</tr>
<tr>
<td>Cons. preference</td>
<td>8.77</td>
<td>60.93</td>
<td>4.23</td>
<td>7.66</td>
<td>19.85</td>
</tr>
<tr>
<td></td>
<td>0.65</td>
<td>30.56</td>
<td>0.10</td>
<td>0.82</td>
<td>12.14</td>
</tr>
<tr>
<td></td>
<td>[0.85]</td>
<td>[83.08]</td>
<td>[2.09]</td>
<td>[67.86]</td>
<td></td>
</tr>
<tr>
<td>Gov. spending</td>
<td>0.00</td>
<td>0.58</td>
<td>53.39</td>
<td>0.00</td>
<td>0.01</td>
</tr>
<tr>
<td></td>
<td>0.00</td>
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<td>[0.01]</td>
<td>[0.00]</td>
<td>[0.02]</td>
<td></td>
</tr>
<tr>
<td>Country premium</td>
<td>22.78</td>
<td>15.07</td>
<td>11.00</td>
<td>81.19</td>
<td>72.05</td>
</tr>
<tr>
<td></td>
<td>21.38</td>
<td>41.43</td>
<td>3.43</td>
<td>86.16</td>
<td>74.12</td>
</tr>
<tr>
<td></td>
<td>[6.62]</td>
<td>[5.54]</td>
<td>[92.92]</td>
<td>[25.81]</td>
<td></td>
</tr>
</tbody>
</table>

Note: The estimates are based on 800,000 draws from the posterior distribution. GPU-FFR estimates –GPU's financial friction model estimated using the mix of Beta and Gamma priors –are in squared brackets.

Table 9: Unconditional variance decomposition (in percentage, %)

Figure 3: Vietnam: Historical variance decomposition of tby
4.4. The importance of the trend shock

AG use the Beveridge - Nelson (BN) decomposition to derive the relative importance of the trend component with respect to the transitory component of the productivity shock in shaping the dynamics of Solow’s residual. The variance ratio reads:

\[
\frac{\sigma_{\Delta \tau}^2}{\sigma_{\Delta sr}^2} = \frac{\alpha^2 \sigma_{\varepsilon}^2}{(1-\rho_g)^2 \sigma_{\Delta sr}^2}
\]

where \(\sigma_{\alpha}^2\) denotes the conventional Solow residuals, and \(\tau_t\) is the trend part of the BN decomposition, such that \(sr_t = \tau_t + s_t\). In Figure 4, we report all relative variance ratios up to lag 12.\(^8\) It appears that, with the exclusion of the Philippines, the trend seems not to be as important as AG predicted when the lag-length increases infinitely.

The above decomposition, however, is not fully convenient since it does not directly contain \(\sigma_{\alpha}\) – the estimated standard deviation of a transitory shock. Since \(a_t\) and \(g_t\) are AR(1) processes, their variances would be \(\sigma_{\alpha}^2/(1-\rho_a^2)\) and \(\sigma_{\varepsilon}^2/(1-\rho_g^2)\), respectively. Recall that \(\sigma_{\alpha}\) and \(\sigma_{\varepsilon}\) are the standard deviations of the corresponding \(\varepsilon_{a,t}\) and \(\varepsilon_{g,t}\), respectively. Thus, the importance of the trend shock can be computed as:\(^9\)

\[
\frac{\text{Var}(g_t)}{\text{Var}(\Delta sr_t)} = \frac{\sigma_{\varepsilon}^2}{(1-\alpha^2)(1+\rho_a)\sigma_{\epsilon}^2 + 2(1-\rho_g^2)\sigma_{\alpha}^2}
\]

with \(\sigma_{\alpha} = 0.68\%\), \(\sigma_{\varepsilon} = 0.54\%\), \(\rho_a = 0.72\), and \(\rho_g = 0.61\). Hence, the relative variance of the trend process \(g_t\) is 28.5%. As \(g_t\) appears in the Cobb-Douglas function with factor \((1-\alpha)\), the model predicts that the non-stationary component only explains \((1-0.35)^2 \times 28.5\% \approx 12\%\) of output movements. This implies that output growth in Vietnam’s economy has been quite stable in response to the non-stationary component of productivity shocks driving Solow’s residual. This estimation is in accordance with the empirical relevance as the ratios displayed in Figure 4.

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\(^8\)We find negligible differences between Cochrane (1988) and AG-modified formula, but the former is easier to compute.

\(^9\)The fact is that \(\text{Var}[(1-\alpha)g_t] = (1-\alpha^2)\sigma_{\varepsilon}^2/(1-\rho_g^2)\) and \(\text{Var}(a_t-a_{t-1}) = 2\text{Var}(a_t) - 2\text{Cov}(a_t,a_{t-1})\).
5. Counterfactual simulations

In this section, we investigate the trajectories of several observables under different loaded shocks. There are some interesting policy implications that arise from analyzing fictitious scenarios and its consequences for Vietnam’s economy. In Scenario 1, transitory productivity shocks, trend shocks and monetary policy shocks are sequentially muted, i.e. $\varepsilon_{t \in \{a,g,\mu\}} = 0$. In Scenario 2, we apply the TFP shocks experienced by Vietnam in 1992 – 1997 to years 2008 – 2013. In Scenario 3, shocks to the country risk premium become silent ($\varepsilon_{\mu,t} = 0$) in 2008 – 2013.
Figure 5: Counterfactual simulation. First row, Panel A and B express Case (1) of setting by turn $\epsilon_a = 0$ (solid line), $\epsilon_g = 0$ (dotted line), and $\epsilon_{\mu} = 0$ (gray dashed line). Second row, Panel C and D correspond to, respectively, the scenarios of setting $\epsilon_{a,2008-2013} = \epsilon_{g,1992-1997}$ (Case 2: solid line), and $\epsilon_{\mu,t} = 0$ for period 2008 - 2013 (Case 3: dotted line). Smoothed series (dashed circle line) depict actual data. Note that the dashed-circle line displays actual data in all panels.

The first row of Figure 5 reports the simulated paths of output growth $g_{y,t}$ (Panel A) and trade-balance-to-output $t_{by,t}$ (Panel B) in the first scenario when we set $\epsilon_a = 0$ (solid line), $\epsilon_g = 0$ (dotted line), and $\epsilon_{\mu} = 0$ (gray dashed line) in every data point in the sample. The smoothed lines (dashed circle line) is the actual data reconstructed using Kalman’s filter. Panels C and D supply analogous information for scenarios 2 and 3.

The counterfactuals in Panels A and B show that in the absence of transitory productivity shocks the trajectory of the trade balance would have stayed unchanged, while the one of income growth would have become more stable. This implies that this kind of shocks help stabilizing the business cycle but have little effect on the external sector. This is in contrast to the impact of trend shocks $\epsilon_{g,t}$ which have little impact on both income growth and the trade balance, as the gap between the actual and simulated lines is small and even non-existent in many data points. With respect to the monetary policy shocks, note that in Panel A the deviation of the gray dashed line becomes more significant in the 2000s reflecting real effects of the proactive Vietnamese monetary policy in the twenty-first century as also found in Anwar and Nguyen (2018). Panel B, however, shows that the monetary policy itself was responsible for the huge trade deficits in the period 2007-2010 despite the fact that Vietnam stabilized over the GFC (remark that in Panel A the positive income growth gap between actual data and the gray dashed line is about 0.7% in 2008 – 2013). All in all, the analysis

\[ \text{In contrast to most western and emerging countries, Vietnam did not have a truly independent central bank for decades (Anwar and Nguyen, 2018). Besides, the State Bank of Vietnam regulated both deposit and lending rates until the beginning of the 2000s (Camen et al., 2006).} \]
in Scenario 1 contradicts, for the case of Vietnam, AG’s claim on the leading role of the trend shock in explaining a sudden drop in trade deficit along with large contractions in connected aggregates.

Panel C shows that if Vietnam’s economy had experienced the same technological improvements as in 1992-1997 – as tested in Scenario 2 –, the simulated path of $g_{yt}$ (solid line) would have evolved above the actual path during the GFC and subsequently. The average gain per year, over 2008 – 2013, amounts to 0.8 percentage points. On the contrary, in the absence of risk premium shocks, the net effect on income growth appears to be small in the same period. These shocks, however, exert a significant influence on the trade deficit as shown by the substantially smaller deficits in 2011-2013 in the absence of such shocks (dotted line in Panel D). The main conclusion we draw from these exercises goes back to the first result. In terms of a high growth and sustainable pace, it seems that technological progress is the critical condition to achieve the best possible path for the Vietnamese economy.

Finally, let us note that our counterfactual simulations strengthen to some degree the findings in Huynh et al. (2017), according to which the monetary policy exerted through interest rates management was insufficient to face the economic downturn. Given our analysis and their evidence based on a calibrated DSGE model, it seems that a strategy more focused on fostering loan supply and targeting productivity generating sectors would have been more successful. We hope this lesson may be helpful in the future design of economic policy in Vietnam.

6. Application to Thailand

We extend our research by applying the analysis to Thailand, which has been, and still is, the key business rival of Vietnam within the ASEAN-5 countries. Before the Asian financial crisis, Thailand was widely recognized as a reference case of an oil-importing emerging economy. For almost 40 years (1958 – 1996), Thailand sustained positive GDP growth rates and achieved “a combination of rapid growth, macroeconomic stability, and steadily declining poverty incidence” (Warr, 2005, p. 4). As explained by the ADB (2015), the success of Thailand was supported by “political stability, a business-friendly regulatory environment, a large domestic market, open access to foreign investment, and greater participation in regional value chains”.

Having faced a debt crisis in 1983 – 1985, Thailand experienced a major boom-bust cycle in the second-half of the 1980s and 1990s, as illustrated in Table 10. Years 1987 – 1996 were characterized by prosperity, with an average of 9% GDP growth per annum that ceased when the region was hit by the 1997 crisis. Thailand was again affected by a series of political and financial shocks,\(^{11}\) but its growth rate kept a pace of 3.97% per year and was classified as an upper-middle-income country in 2011 (World Bank, 2011). As Table 10 shows, in spite of this success, Vietnam’s economy has outperformed Thailand’s one by approximately two percentage points since 2000.

\(^{11}\)These were a coup d’état in 2006 (political shock), the global 2008 crisis (financial shock), and the flood in 2011 (economic shock).

<table>
<thead>
<tr>
<th>Period</th>
<th>rGDP growth</th>
<th>rGDPpc growth</th>
<th>rGDP growth</th>
<th>rGDPpc growth</th>
</tr>
</thead>
<tbody>
<tr>
<td>1970 – 1986, Pre-boom (*)</td>
<td>6.08</td>
<td>3.82</td>
<td>4.67</td>
<td>2.39</td>
</tr>
<tr>
<td>1987 – 1996, Boom (**)</td>
<td>9.00</td>
<td>7.78</td>
<td>6.54</td>
<td>4.53</td>
</tr>
<tr>
<td>2008 – 2009, Global-crisis</td>
<td>0.48</td>
<td>0.34</td>
<td>5.38</td>
<td>4.41</td>
</tr>
<tr>
<td>2010 – 2015, Post-crisis</td>
<td>3.55</td>
<td>3.20</td>
<td>5.82</td>
<td>4.72</td>
</tr>
</tbody>
</table>

(*) and (**) show Thailand Pre-boom and Boom periods as documented in Warr (2005, p.5).

rGDP (rGDPpc) is the rate of growth of real GDP (per capita).

Table 11: Thailand: Bayesian estimation, 1976 – 2015

<table>
<thead>
<tr>
<th>Params</th>
<th>$\sigma_{gy}$</th>
<th>$\sigma_{gcp}/\sigma_{gy}$</th>
<th>$\sigma_{cg}/\sigma_{gy}$</th>
<th>$\sigma_{inv}/\sigma_{gy}$</th>
<th>$\sigma_{tby}/\sigma_{gy}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model</td>
<td>3.55</td>
<td>1.39</td>
<td>1.78</td>
<td>6.95</td>
<td>4.60</td>
</tr>
<tr>
<td>Data</td>
<td>3.62</td>
<td>1.12</td>
<td>1.07</td>
<td>3.71</td>
<td>2.15</td>
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<td></td>
<td>(0.088)</td>
<td>(0.238)</td>
<td>(0.419)</td>
<td>(0.366)</td>
<td></td>
</tr>
</tbody>
</table>

$\sigma_a$, $\sigma_g$, $\sigma_p$, $\sigma_{cg}$, $\sigma_{g}$, $\sigma_{tby}$, $\sigma_{inv}$, $\sigma_{gy}$

$\rho_a$, $\rho_g$, $\rho_p$, $\rho_{cg}$, $\rho_{tby}$, $\rho_{inv}$

$\psi$, $\gamma$, $\phi$, $\beta^*$, $\mu_g$

| Standard error in parentheses; $\sigma$ denotes standard deviation. |

$g_y$, $g_{cp}$, $g_{cg}$, $g_{inv}$ and $tby$ variables are defined as in section 5.

The bottom block of Table 11 shows that the long-term growth rate of Thailand is lower than that of Vietnam by 0.8 percentage points, but the subjective discount rate of Thailand is just one-half of the Vietnamese

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Except for $\omega$ and $\theta$, the same calibration set is used to bring the model to Thailand data. Since Thailand is calibrated in AG, we reuse $\omega = 1.6$ so that the elasticity of labor supply, $\frac{1}{(\omega - 1)}$, is set at 1.7 and $\theta = 1.4$. Table 11 demonstrates that the excess volatility of the growth variables with respect to output is essentially matched. The predicted parameter $\sigma_{gy}$ is close to the actual data, and the downward slope of $tby$’s autocorrelation function has been exactly reproduced, as plotted in Figure 2. Nevertheless, the relative standard deviations are overly predicted because the standard model could not handle properly the two structural breaks (or debt crises) experienced by Thailand’s economy in 1982 – 1985 and 1997 – 1999. As in the case of Vietnam, several growth variables’ contemporaneous and first-order autocorrelation coefficients are understated by the model.\(^{12}\)

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\(^{12}\)Not reported in the main text since the estimates are similar to the ones for Vietnam.
counterpart. The latter implies that Thailand’s economy is better structured even though capital adjustment costs, $\phi$, are similar. The lower temporal discount rate reflects the stronger attachment of Thailand’s households to their lifetime income. In support to this claim is the low estimate of the habit persistence coefficient, $\gamma = 0.13$. In addition, the estimated parameter $\psi$, which controls for the sensitivity of the country risk premium, is one-third of Vietnam’s counterpart. This may help to understand Thailand’s long-lasting trade balance surplus in the aftermath of the 1997 crisis.

The long-term variance decomposition Table 12 proves that the explanatory power of both the transitory and trend shocks is relatively weak since neither of them can account for more than 27% of the output growth variance. In particular, the relevance of the non-stationary component is as low as 6.18%. On the other hand, the country risk premium is vital to explain Thailand’s economy developments, as it accounts for 54.6% of the unconditional variance over the sample range.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Output growth</th>
<th>Consumption growth</th>
<th>Gov spending growth</th>
<th>Investment growth</th>
<th>Trade-balance to GDP ratio</th>
</tr>
</thead>
<tbody>
<tr>
<td>Nonstationary tech</td>
<td>17.59</td>
<td>1.97</td>
<td>4.84</td>
<td>5.24</td>
<td>2.55</td>
</tr>
<tr>
<td>Stationary tech</td>
<td>26.74</td>
<td>4.86</td>
<td>7.25</td>
<td>0.69</td>
<td>0.78</td>
</tr>
<tr>
<td>Preference shift</td>
<td>0.05</td>
<td>5.80</td>
<td>0.01</td>
<td>0.05</td>
<td>0.51</td>
</tr>
<tr>
<td>Cons. preference</td>
<td>1.01</td>
<td>21.83</td>
<td>0.28</td>
<td>0.41</td>
<td>3.48</td>
</tr>
<tr>
<td>Gov. spending</td>
<td>0.00</td>
<td>4.82</td>
<td>71.31</td>
<td>0.00</td>
<td>0.01</td>
</tr>
<tr>
<td>Country premium</td>
<td>54.62</td>
<td>60.71</td>
<td>16.30</td>
<td>93.60</td>
<td>92.67</td>
</tr>
</tbody>
</table>

Note: The estimates are based on 800,000 draws from the posterior distribution.


7. Concluding remarks

This paper provides a detailed analysis of Vietnam’s economy in connection to its ASEAN-5 peers. In the first part, we provide information on two complementary sides of Vietnam’s performance in terms of growth sources and business cycle drivers. In the second part, we develop a DSGE-RBC model. Our model departs from those in AG and GPU, but further incorporates habit formation. In this way it provides a close match of the facts in which our analysis of Vietnam, and the comparison with Thailand, can be conducted.

We show that the contribution of TFP to Vietnam’s economic growth is approximately one-third on average in 1986–2015, although it drops to less than 20% in the 2000s. In turn, capital accumulation has driven Vietnam’s economy since 1992, in parallel to the acceleration in the opening and deregulation processes that started that year. One of the main consequences of such processes was the development of a structural trade deficit.

In terms of business cycle characteristics, we document a similar behavior in Vietnam than in the ASEAN-5 economies, which does not differ significantly from the one in other emerging market economies.
The intrinsic difficulties of RBC models in replicating short-term observational dynamics is well-known. With this caveat in mind, it is important to note that our model provides a better account of the facts in Vietnam and Thailand than other reference models in the emerging markets literature—for example, the ones by AG and GPU, which provide the departure point of our modelling strategy.

Provided with this improved setting, the variance decomposition analysis reveals that transitory productivity shocks account for around 50% of Vietnam’s output growth fluctuations. Country-risk premium shocks are also relevant, although they are far more critical in the case of Thailand, as they totally dominate the impact of the productivity shock throughout the whole period (1976–2015). Given these findings, we conclude that Thailand’s economy is more vulnerable to international externalities than Vietnam’s one, which has stricter capital flow controls and is still, de-facto, a non-free market economy. It is probably on account of Vietnam’s uncompleted transition that technological progress and productivity-enhancing measures come out as fundamental to secure a sustainable high growth path. All other examined growth drivers seem to be of secondary order.

Another crucial finding is the scarce contribution of non-stationary TFP shocks to Solow’s residual volatility, 12% in Vietnam and 6% in Thailand. These results, which refute AG’s claim that “the cycle is the trend” in emerging market economies, add-up to recent literature with similar results for Argentina, Mexico and Korea.

References
Huynh, P., Nguyen, T., Duong, T., Pham, D., 2017. Leaning against the wind policies on vietnam’s economy with dsge model. Economies 5 (1), 3.
URL https://ideas.repec.org/b/wbk/wbpubs/23724.html
Figure A1: Priors and Posteriors
Figure A2: Priors and Posteriors (in the case of estimated $\pi$ with Beta prior)

Subjective discount rate, $\beta^*$
Long-run growth rate, $\mu_g$
Habit persistence, $\gamma$
Capital adjustment costs, $\phi$
Debt elasticity, $\psi$

Figure A3: Selected trace plots
Figure A4: Elementary Effects analysis (in the case of estimated $\pi$ with Beta(0,1) prior)

Figure A5: Sensitivity analysis w.r.t $g_y$ and $g_{cp}$ (in the case of estimated $\pi$). The higher bar indicates the stronger effects