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What drives fluctuations of labor wedge and business cycles? Evidence from Japan *

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August 31, 2020

^{*}This work is financially supported by the Zengin Foundation for Studies on Economics and Finance in 2018 and the Research Institute for Socionetwork Strategies of Kansai University in 2020.

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Abstract

It is well known that the labor wedge worsens in the recessions. The main research questions of this study are as follows. (i) What is the main driving force of the labor wedge? (ii) Is the main driver of the labor wedge the same as that of business cycles? In this study, we analyze which structural shocks drive the fluctuation of the labor wedge and business cycles using a canonical medium-scale dynamic stochastic general equilibrium model with nominal and real frictions. The model is estimated using Japanese data. One of the novel features of this study is our estimation strategy. In standard Bayesian estimation, the prior distribution of the parameters for the standard deviations of the structural shocks is inverse gamma, which does not support zero value and assumes the existence of structural shocks. On the contrary, we employ a more flexible prior distribution of the parameters for the standard deviations of structural shocks to allow the non-existence of structural shocks. Under the standard prior distribution, the estimation results imply that the labor wedge is mainly driven by preference and transitory technology shocks, whereas the investment adjustment cost shock is the most important for the business cycle fluctuations. However, under our relaxed prior distribution, which allows the non-existence of structural shocks, the estimation results imply that both the labor wedge and the business cycles are mainly driven by permanent technology and investment adjustment cost shocks.

Keywords: Labor wedge; DSGE model; structural shocks; measurement error; prior distribution

JEL codes: E32; E37

1 Introduction

It is well known that the labor wedge, defined as the gap between the marginal rate of substitution and the marginal productivity of labor, worsens in recessions, including the Great Depression and Great Recession. The labor wedge has been investigated as an important variable to understand business cycle fluctuations.

The main research questions of this study are as follows. (i) What is the main driving force of the labor wedge in a canonical medium-scale dynamic stochastic general equilibrium (DSGE) economy? (ii) Is the main driver of the labor wedge the same as that of business cycles? The recent literature on the medium-scale DSGE model has found that such a model can account for the salient aspects of business cycle fluctuations. In the literature, many works investigate the main driving force of output fluctuations, whereas there are few studies on the source of the fluctuations of the labor wedge.¹ If the labor wedge is an important variable to understand business cycles, then the main driving force of the labor wedge should be the same as that of business cycles.

In this study, we analyze which structural shocks drive the fluctuation of the labor wedge and business cycles using a canonical medium-scale DSGE model. There are many nominal and real frictions, and many structural shocks in the model. We estimate the model using Japanese data. One of special features of this study is our estimation strategy. In the standard Bayesian estimation of DSGE models, the prior distribution of the parameters for the standard deviations of the shocks is the inverse gamma distribution, which does not support zero value and assumes the existence of structural shocks. On the contrary, in this study, we employ a more flexible prior distribution of the parameters for the standard deviations of shocks and measurement errors to allow the non-existence of shocks and measurement errors. Following Ferroni, Grassi, and

¹An exception is Cheremukhin and Restrepo-Echavarria (2014), who investigate the source of the labor wedge in a DSGE model with search and matching in the labor market, while our model is a canonical medium-scale DSGE model without search and matching frictions.

Leon-Ledesma (2019), we employ normal distribution in this study.

Under standard prior distribution of the parameters for the standard deviations of structural shocks, the estimation results imply that the labor wedge is mainly driven by preference and transitory technology shocks, whereas the investment adjustment cost shock is the most important for business cycle fluctuations. Meanwhile, under our relaxed prior distribution, which allows the non-existence of shocks, the estimation results show that both the labor wedge and the business cycles are mainly driven by permanent technology and investment adjustment cost shocks.

Our estimation results imply the followings. If our estimation were to employ inverse gamma distribution, which is standard in the literature but assumes the existence of structural shocks, then the investigation of the source of the labor wedge fluctuation might not be promising to understand business cycles. This is because the source of the labor wedge is different from that of business cycles. However, by using our more relaxed prior, the normal distribution, to allow the non-existence of structural shocks, the investigation of the labor wedge would be promising to understand business cycles: both the labor wedge and business cycles are driven by the same structural shocks.

Related literature: The importance of the labor wedge in business cycles is emphasized by Chari, Kehoe, and McGrattan (2002, 2007), Kobayashi and Inaba (2006), Shimer (2009), Ohanian (2010), and Otsu (2011). Chari, Kehoe, and McGrattan (2002, 2007) propose the business cycle accounting method and find that the labor wedge is important for the US Great Depression and the 1982 recession. Kobayashi and Inaba (2006) find that the labor wedge is helpful to account for the Japan's lost decade during the 1990s and the recession during the 1920s. Shimer (2009) and Ohanian (2010) find that the labor wedge worsens during recession. In particular, Ohanian (2010) focuses on the US Great Recession. Otsu (2011) investigates the dynamics of the labor wedge of Japanese economy from 1980 to 2007. The main driving force of the fluctuation of the labor wedge is investigated by Hall (2009), Shimer (2009), Pescatori and Tasci (2011), Cheremukhin and Restrepo-Echavarria (2014), Karabarbounis (2014a, 2014b), Duras (2017), and Zhang (2018). Hall (2009), Shimer (2009), Pescatori and Tasci (2011), and Cheremukhin and Restrepo-Echavarria (2014) emphasize the role of matching frictions in the labor market. Duras (2017) finds that matching frictions in the goods market, in addition to the labor market, help to account for fluctuations of the labor wedge.

Karabarbounis (2014a) empirically finds that fluctuations of the labor wedge mainly reflect fluctuations of the gap between the real wage and the marginal rate of substitution. He emphasizes the importance of modeling the household side of the labor market. Karabarbounis (2014b) proposes that home production in the utility function accounts for the international findings of the labor wedge. Zhang (2018) finds that the collateral constraints of entrepreneurs is helpful to account for the variation in the labor wedge during the credit crunch.

Our medium-scale DSGE model does not include matching frictions, home production, and collateral constraints, as investigated by these existing works. In our model, habit persistence and sticky wage are considered, generating a gap between the marginal rate of substitution and the real wage. This setup is consistent with the empirical finding of Karabarbounis (2014a).

The sources of business cycle fluctuations have been investigated by many researchers. In particular, the importance of technology shock is emphasized traditionally as in King and Rebelo (1999) and Hayashi and Prescott (2002). Recent works by Kaihatsu and Kurozumi (2014a, 2014b) find that technology shocks are the main driving force of business cycles both in the US and Japan. Meanwhile, Justiniano, Primiceri, and Tambalotti (2010, 2011) and Hirose and Kurozumi (2012) emphasize the importance of the investment adjustment cost shock in business cycles. In our estimate, the (permanent) technology and investment adjustment cost shocks are important for business cycles,

which are consistent with these findings in the previous research.

Our estimation method is based on Ferroni, Grassi, and Leon-Ledesma (2019), who propose an estimation method allowing the non-existence of structural shocks. The authors find that government spending, price markup, and wage markup shocks do not generate significant dynamics using the model of Smets and Wouters (2007) in the US.

The remainder of the paper is organized as follows. Section 2 introduces the model. Section 3 shows the data and the estimation strategy of the model. Section 4 explains the main results. Finally, Section 5 concludes.

2 The Model

2.1 The Model

The model is a variant of the medium-scale DSGE model à la Christiano, Eichenbaum, and Evans (2005) and Smets and Wouters (2007). There are many real and nominal frictions to account for the salient aspects of business cycles. As real frictions, habit persistence, adjustment costs of investment, and variable capital utilization are introduced. As nominal frictions, Calvo-type nominal price and wage rigidities with partial inflation indexations are introduced. The central bank follows a Taylor-type nominal interest rate rule. There are many structural shocks: permanent and transitory technology, preference (a shock to the discount factor), labor supply (a shock to the weight of disutility from labor supply), investment adjustment cost, price markup, wage markup, government purchases, and monetary policy shocks.²³

²In our model, the investment specific technology shock is eliminated from the model of Hirose and Kurozumi (2012), and the transitory technology shock is introduced.

³As explained in section 2.2, we redefine the wage shock as the linear combination of the labor supply and wage markup shocks, because the labor supply and wage markup shocks cannot be identified in this model.

Final-good firms: The final-good firms are perfectly competitive. They produce homogeneous final-good Y_t using intermediate-good $Y_t(f)$. The production function is given by

$$Y_t = \left[\int_0^1 Y_t(f)^{\frac{1}{1+\lambda_t^p}} df\right]^{1+\lambda_t^p},\tag{1}$$

where λ_t^p is a time-varying parameter for elasticity of substitution among intermediategood $\theta_t^p > 1$, which is defined by $\lambda_t^p = 1/(\theta_t^p - 1) > 0$. Profit maximization implies the demand function of intermediate-good $Y_t(f)$:

$$Y_t(f) = \left[\frac{P_t(f)}{P_t}\right]^{-\frac{1+\lambda_t^P}{\lambda_t^P}} Y_t,$$
(2)

where P_t is the price of final-good Y_t , and $P_t(f)$ is the price of intermediate-good $Y_t(f)$.

Intermediate-good firms: The intermediate-good firms are monopolistically competitive. The intermediate-good firm indexed by $f \in [0, 1]$ produces differentiated intermediate-good $Y_t(f)$ using labor input $\ell_t(f)$ and capital service $K_t^S(f)$. The production function is given by

$$Y_t(f) = \exp(z_t^{zt}) [K_t^S(f)]^{\alpha} [Z_t^P \ell_t(f)]^{1-\alpha} - \Phi Z_t^P,$$
(3)

where $\alpha \in (0, 1)$ is the cost share of capital; Φ is the fixed cost of production; z_t^{zt} is the transitory technology shock; and Z_t^P is the permanent technology that evolves according to

$$\log Z_t^p = z^* + \log Z_{t-1}^P + z_t^{zp}, \tag{4}$$

where z^* is the steady-state growth rate of Z_t^P , and z_t^{zp} is the permanent technology shock. Both z_t^{zt} and z_t^{zp} follow the AR(1) process. The last term in the production function, ΦZ_t^P , is multiplied by Z_t^P to guarantee the existence of the balanced growth path. The cost minimization of intermediate-good firms implies

$$R_t^k = mc_t \,\alpha \exp(z_t^{zt}) \left[\frac{Z_t^P \ell_t(f)}{K_t^S(f)} \right]^{1-\alpha},\tag{5}$$

$$W_t = mc_t (1 - \alpha) \exp(z_t^{zt}) Z_t^P \left[\frac{Z_t^P \ell_t(f)}{K_t^S(f)} \right]^{-\alpha},$$
(6)

where mc_t is the real marginal cost; R_t^k is the rental rate of capital; and W_t is the real wage rate.

Calvo-type sticky prices are introduced. In every period, a fraction $\xi_p \in [0, 1]$ of intermediate-good firms can reoptimize their prices. The other firms index their prices to the weighted average of past inflation (π_{t-1}) and the steady-state inflation (π^*) : $\pi_{t-1}^{\gamma_p}(\pi^*)^{1-\gamma_p}$, where $\gamma_p \in [0, 1]$ is the relative weight of the past inflation. The objective function of the intermediate-good firms that reoptimize their prices at period *t* is

$$E_t \sum_{j=0}^{\infty} \left(\beta \xi_p\right)^j \left(\frac{\Lambda_{t+j}}{\Lambda_t}\right) \left[\frac{P_t(f)}{P_{t+j}} \prod_{k=1}^j \left(\pi_{t+k-1}^{\gamma_p}(\pi^*)^{1-\gamma_p}\right) - mc_{t+j}\right] Y_{t+j}(j), \tag{7}$$

where Λ_t is the marginal utility of consumption of households, and $\beta^j \frac{\Lambda_{t+j}}{\Lambda_t}$ is the stochastic discount factor. The demand function for $Y_{t+j}(f)$ is given by

$$Y_{t+j}(f) = \left[\frac{P_t(f)}{P_{t+j}} \prod_{k=1}^j \left(\pi_{t+k-1}^{\gamma_p}(\pi^*)^{1-\gamma_p}\right)\right]^{-\frac{1+\lambda_{t+j}^p}{\lambda_{t+j}^p}} Y_{t+j}.$$
(8)

The reoptimized price P_t^o is the same for all intermediate-good firms. The first-order condition for reoptimized price P_t^o is

$$1 = \frac{E_t \sum_{j=0}^{\infty} (\beta \xi_p)^j \frac{(1+\lambda_{t+j}^p) m c_{t+j} \Lambda_{t+j} Y_{t+j}}{\lambda_{t+j}^p} \left[\frac{P_t^0}{P_t} \prod_{k=1}^j \left(\frac{\pi_{t+k-1}}{\pi^*} \right)^{\gamma_p} \frac{\pi^*}{\pi_{t+k}} \right]^{-\frac{1+\lambda_{t+j}^p}{\lambda_{t+j}^p}}}{E_t \sum_{j=0}^{\infty} (\beta \xi_p)^j \frac{\Lambda_{t+j} Y_{t+j}}{\lambda_{t+j}^p} \left[\frac{P_t^0}{P_t} \prod_{k=1}^j \left(\frac{\pi_{t+k-1}}{\pi^*} \right)^{\gamma_p} \frac{\pi^*}{\pi_{t+k}} \right]^{-\frac{1}{\lambda_{t+j}^p}}}.$$
(9)

Households: The household indexed by $h \in [0, 1]$ consumes $C_t(h)$, invests $I_t(h)$, holds safe asset $B_t(h)$ and capital stock $K_t(h)$, and supplies differentiated labor service $\ell_t(h)$.

The utility function is given by

$$E_0 \sum_{t=0}^{\infty} \beta^t \exp(z_t^b) \left[\frac{[C_t(h) - \theta C_{t-1}(h)]^{1-\sigma}}{1-\sigma} - \frac{Z_t^{1-\sigma} \exp(z_t^\ell) \ell_t(h)^{1+\chi}}{1+\chi} \right],$$
(10)

where $\beta \in (0, 1)$ denotes the discount factor; $\sigma > 0$ is the elasticity of intertemporal substitution; $\theta \in (0, 1)$ is the degree of habit persistence; $\chi > 0$ denotes the inverse of the labor supply elasticity; and z_t^b and z_t^ℓ are the structural shocks to discount factor (preference shock) and labor supply (labor supply shock), respectively. The technology term $Z_t^{1-\sigma}$ appears in the disutility of labor supply, as employed by Erceg, Guerrieri, and Gust (2006), to ensure the existence of a balanced growth path.

The budget constraint of the household is

$$C_{t}(h) + I_{t}(h) + \frac{B_{t}(h)}{P_{t}} = W_{t}(h)\ell_{t}(h) + R_{t}^{k}u_{t}(h)K_{t-1}(h) + \frac{R_{t-1}^{n}B_{t-1}(h)}{P_{t}} + T_{t}(h),$$
(11)

where P_t denotes the price level; $W_t(h)$ is the real wage rate; R_t^n is the nominal gross interest rate; R_t^k denotes the real rental rate of capital; $u_t(h)$ is the capital utilization rate; $K_{t-1}(h)$ is capital stock at the end of period t - 1; and $T_t(h)$ denotes a transfer from the government and firms.

The capital stock evolves as follows:

$$K_t(h) = [1 - \delta(u_t(h))]K_{t-1}(h) + \left[1 - S\left(\frac{I_t(h)}{I_{t-1}(h)}\frac{\exp(z_t^i)}{z^*}\right)\right]I_t(h),$$
(12)

where z_t^i is the investment adjustment cost shock. In this specification, the cost of high capital utilization is a high depreciation rate of capital as in Greenwood, Hercowitz, and Huffman (1988); $\delta'(\cdot) > 0$, $\delta''(\cdot) > 0$, and $\delta'(u^*)/\delta''(u^*) = \mu$, where u^* is the steady-state capital utilization rate. The functional form of the adjustment costs of investment is given by

$$S(x) = \frac{1}{\zeta} \left(x - 1 \right)^2,$$
 (13)

where $1/\zeta > 0$ is the degree of adjustment cost of investment.

Because of the existence of complete insurance markets, the decisions of $C_t(h)$, $B_t(h)$, $u_t(h)$, $K_t(h)$, and $I_t(h)$ are the same for all households; then, the first-order conditions are given by

$$\Lambda_t = \exp(z_t^b) \left[C_t - \theta C_{t-1} \right]^{-\sigma} - \beta \theta E_t \left(\exp(z_{t+1}^b) \left[C_{t+1} - \theta C_t \right] \right), \tag{14}$$

$$\Lambda_t = \beta E_t \left[\Lambda_{t+1} \cdot \frac{\kappa_t}{\pi_{t+1}} \right], \tag{15}$$

6)

$$R_t^k = Q_t \,\delta'(u_t),\tag{1}$$

$$1 = Q_{t} \left[1 - S \left(\frac{I_{t}}{I_{t-1}} \frac{\exp(z_{t}^{i})}{z^{*}} \right) - S' \left(\frac{I_{t}}{I_{t-1}} \frac{\exp(z_{t}^{i})}{z^{*}} \right) \frac{I_{t}}{I_{t-1}} \frac{\exp(z_{t}^{i})}{z^{*}} \right] + \beta E_{t} \left[\frac{\Lambda_{t+1}}{\Lambda_{t}} Q_{t+1} S' \left(\frac{I_{t+1}}{I_{t}} \frac{\exp(z_{t+1}^{i})}{z^{*}} \right) \left(\frac{I_{t+1}}{I_{t}} \right)^{2} \frac{\exp(z_{t+1}^{i})}{z^{*}} \right], \quad (17)$$

$$Q_{t} = \beta E_{t} \left[\frac{\Lambda_{t+1}}{\Lambda_{t}} \left\{ R_{t+1}^{k} u_{t+1} + Q_{t+1} \left(1 - \delta(u_{t+1}) \right) \right\} \right],$$
(18)

where Λ_t is the marginal utility of consumption; $\pi_{t+1} = P_{t+1}/P_t$ is the gross price inflation rate; and Q_t is the real price of capital, which is defined as the ratio of the Lagrange multiplier of the evolution of capital to the marginal utility Λ_t .

Wage setting: The household supplies its differentiated labor service $\ell_t(h)$ to the intermediategood firms. Then, the labor market is monopolistically competitive. The intermediategood firm *f* aggregates its labor inputs $\ell_t(f, h)$ according to the following technology:

$$\ell_t(f) = \left[\int_0^1 \ell_t(f,h)^{\frac{1}{1+\lambda_t^w}} dh\right]^{1+\lambda_t^w},\tag{19}$$

where λ_t^w is a time-varying parameter for the elasticity of substitution between labor supplies $\theta_t^w > 1$, which is defined by $\lambda_t^w = 1/(\theta_t^w - 1) > 0$. The cost minimization of the intermediate-good firm and the aggregation over intermediate-good firms imply the following demand function of labor $\ell_t(h)$:

$$\ell_t(h) = \left[\frac{W_t(h)}{W_t}\right]^{-\frac{1+\lambda_t^w}{\lambda_t^w}} \ell_t.$$
(20)

Calvo-type sticky wages are introduced as in Erceg, Henderson, and Levin (2000). In every period, a fraction $\xi_w \in [0, 1]$ of households can reoptimize their wages. The other households index their wages to both the gross steady-state balanced growth rate (z^*) and the weighted average of past inflation and the steady-state inflation : $\pi_{t-1}^{\gamma_w}(\pi^*)^{1-\gamma_w}$, where $\gamma_w \in [0, 1]$ is the relative weight of the past inflation. The objective function is

$$E_{t} \sum_{j=0}^{\infty} \left(\beta \xi_{w}\right)^{j} \left[\Lambda_{t+j} \ell_{t+j}(h) \left(\frac{P_{t} W_{t}(h)}{P_{t+j}} \right) \prod_{k=1}^{j} \left(z^{*} \pi_{t+k-1}^{\gamma_{w}}(\pi^{*})^{1-\gamma_{w}} \right) - \frac{Z_{t+j}^{1-\sigma} \exp(z_{t+j}^{\ell}) \ell_{t+j}(h)^{1+\chi}}{1+\chi} \right],$$
(21)

and the labor demand function is given by

$$\ell_{t+j}(h) = \left[\frac{P_t W_t(h)}{P_{t+j} W_{t+j}} \prod_{k=1}^j \left(z^* \pi_{t+k-1}^{\gamma_w} (\pi^*)^{1-\gamma_w}\right)\right]^{-\frac{1+\lambda_t^w}{\lambda_t^w}} \ell_{t+j}.$$
 (22)

The reoptimized wage W_t^o is the same for all households. The first-order condition for reoptimized wage W_t^o is

$$1 = \frac{E_{t} \sum_{j=0}^{\infty} (\beta \xi_{w})^{j} \frac{(1+\lambda_{t+j}^{w}) \exp(z_{t+j}^{b}) \exp(z_{t+j}^{\ell}) Z_{t+j}^{1-\sigma}}{\lambda_{t+j}^{w}} \left(\ell_{t+j} \left[\frac{W_{t}^{0}(z^{*})^{j}}{W_{t+j}} \prod_{k=1}^{j} \left(\frac{\pi_{t+k-1}}{\pi^{*}} \right)^{\gamma_{w}} \frac{\pi^{*}}{\pi_{t+k}} \right]^{-\frac{1+\lambda_{t+j}^{w}}{\lambda_{t+j}^{w}}} \right)^{1+\chi}}{E_{t} \sum_{j=0}^{\infty} (\beta \xi_{w})^{j} \frac{\Lambda_{t+j} W_{t+j}}{\lambda_{t+j}^{w}}} \ell_{t+j} \left[\frac{W_{t}^{o}(z^{*})^{j}}{W_{t}} \prod_{k=1}^{j} \left(\frac{\pi_{t+k-1}}{\pi^{*}} \right)^{\gamma_{w}} \frac{\pi^{*}}{\pi_{t+k}} \right]^{-\frac{1}{\lambda_{t+j}^{w}}}.$$
(23)

Central bank: The central bank follows a Taylor-type nominal interest rate rule:

$$\log R_t^n = \phi_r \log R_{t-1}^n + (1 - \phi_r) \left[\log \bar{R}^n + \phi_\pi \frac{1}{4} \sum_{j=0}^3 \log \frac{\pi_{t-j}}{\pi^*} + \phi_y \log \frac{Y_t}{Y_t^P} \right] + z_t^r, \quad (24)$$

where π^* is the steady-state inflation rate; Y_t^P is the potential output; and z_t^r is the monetary policy shock. Parameter $\phi_r \in [0, 1)$ represents the degree of interest rate smoothing, and $\phi_{\pi} > 1$ and $\phi_y \ge 0$ are the monetary policy responses to inflation and output, respectively. The potential output Y_t^P is defined by

$$Y_t^P = [u^* k^* Z_{t-1}^P]^{\alpha} [Z_t^P \ell^*]^{1-\alpha} - \Phi Z_t^P,$$
(25)

where u^* is the steady-state of capital utilization; k^* is the steady-state detrended capital stock (K_t/Z_t^P) ; and ℓ^* is the steady-state hours worked. This specification of the potential output is similar to the estimates of Hara et al. (2006). In this specification, only the permanent (technology) shock is considered as a driving force of potential output. This setup is similar to the estimates of Fueki et al. (2016).

Aggregations and market clearing conditions: Since the decisions on $u_t(h)$, $I_t(h)$, and $K_t(h)$ are the same for all households, the evolution of the capital stock (12) is given by

$$K_{t} = [1 - \delta(u_{t})]K_{t-1} + \left[1 - S\left(\frac{I_{t}}{I_{t-1}}\frac{\exp(z_{t}^{i})}{z^{*}}\right)\right]I_{t}.$$
(26)

The capital market-clearing conditions are given by

$$u_t K_{t-1} = \int_0^1 K_t^S(f) df.$$
 (27)

Combining the cost-minimization conditions of intermediate-good firms (5) and (6) and aggregation over intermediate-good firms yields

$$\frac{1-\alpha}{\alpha} = \frac{W_t \ell_t}{R_t^k u_t K_{t-1}}.$$
(28)

The real marginal cost is given by

$$mc_t = \frac{1}{\exp(z_t^{zt})} \left(\frac{W_t}{(1-\alpha)Z_t^P}\right)^{1-\alpha} \left(\frac{R_t^k}{\alpha}\right)^{\alpha}.$$
(29)

Aggregating the production function (3) over intermediate-good firms yields

$$Y_t s_t = \exp(z_t^{zt}) \left[u_t K_{t-1} \right]^{\alpha} \left[Z_t^{\mathcal{P}} \ell_t \right]^{1-\alpha} - \Phi Z_t^{\mathcal{P}}, \tag{30}$$

where $s_t = \int_0^1 [P_t(f)/P_t]^{-(1+\lambda_t^p)/\lambda_t^p} df$ is the price dispersion of the intermediate-good price. This price dispersion can be ignored in the linearized system around the steady state where the steady-state value is one.

Finally, the resource constraint is

$$C_t + I_t + g^* Z_t^P \exp(z_t^g) = Y_t,$$
 (31)

where g^* is the steady-state ratio of government purchases to output; and z_t^g is a government shock.

Labor wedge: Following Shimer (2009), the labor wedge is defined as follows.

(Labor wedge) =
$$\frac{1}{1-\alpha} \left[\frac{C_t}{Y_t} \right] \ell_t^{\frac{1+\chi}{\chi}}$$
. (32)

This specification is based on the period utility $\log(C_t(h)) - \frac{\ell_t(h)^{1+\chi}}{1+\chi}$, and the Cobb–Douglas production function.

2.2 Log-Linearized Equilibrium Conditions

The endogenous variables except for marginal utility Λ_t are detrended by the technology level Z_t as $x_t = X_t/Z_t$. The marginal utility Λ_t is detrended as $\lambda_t = \Lambda_t/Z_t^{-\sigma}$. The equilibrium conditions are log-linearized around a steady state. The variable with tilde \tilde{x}_t is defined as the log deviation of x_t from its steady-state value x^* :

$$\tilde{x}_t = \log(x_t) - \log(x^*).$$

The log-linearized equilibrium system is described as follows. The marginal utility of consumption (14) is

$$\left(1 - \frac{\theta}{z^*}\right) \left(1 - \frac{\beta\theta}{(z^*)^{\sigma}}\right) \tilde{\lambda}_t = -\sigma \left\{ \tilde{c}_t - \frac{\theta}{z^*} (\tilde{c}_{t-1} - z_t^z) \right\} + \left(1 - \frac{\theta}{z^*}\right) z_t^b$$

$$+ \frac{\beta\theta}{(z^*)^{\sigma}} \left[\sigma \left\{ E_t \tilde{c}_{t+1} + E_t z_{t+1}^z - \frac{\theta}{z^*} \tilde{c}_t \right\} - \left(1 - \frac{\theta}{z^*}\right) E_t z_{t+1}^b \right].$$
(33)

The Euler equation for nominal bond (15) is

$$\tilde{\lambda}_t = E_t \tilde{\lambda}_{t+1} - \sigma E_t z_{t+1}^z + \tilde{R}_t^n - E_t \tilde{\pi}_{t+1}.$$
(34)

The capital utilization (16) is

$$\tilde{u}_t = \mu(\tilde{R}_t^k - \tilde{q}_t). \tag{35}$$

The Euler equation for investment (17) is

$$\frac{1}{\zeta} \left\{ \tilde{i}_t - \tilde{i}_{t-1} + z_t^z + z_t^i \right\} = \tilde{q}_t + \frac{\beta(z^*)^{1-\sigma}}{\zeta} \left\{ E_t \tilde{i}_{t+1} - \tilde{i}_t + E_t z_{t+1}^z + E_t z_{t+1}^i \right\}.$$
(36)

The Euler equation for capital (18) is

$$\tilde{q}_{t} = E_{t}\tilde{\lambda}_{t+1} - \lambda_{t} - \sigma E_{t}z_{t+1}^{z} + \frac{\beta}{(z^{*})^{\sigma}} \left\{ R^{k}E_{t}\tilde{R}_{t+1}^{k} + (1-\delta)E_{t}\tilde{q}_{t+1} \right\}.$$
(37)

The New Keynesian Phillips curve (9) is

$$\tilde{\pi}_{t} - \gamma_{p} \tilde{\pi}_{t-1} = \beta(z^{*})^{1-\sigma} (E_{t} \tilde{\pi}_{t+1} - \gamma_{p} \tilde{\pi}_{t}) + \frac{(1 - \xi_{p})(1 - \beta \xi_{p} z^{1-\sigma})}{\xi_{p}} \widetilde{mc}_{t} + z_{t}^{p}.$$
(38)

where z_t^p denotes the price markup shock, which is defined by

$$z_t^p = \frac{(1-\xi_p)(1-\beta\xi_p\bar{z}^{1-\sigma})}{\xi_p}\tilde{\lambda}_t^p.$$

The New Keynesian wage Phillips curve (23) is

$$\widetilde{w}_{t} - \widetilde{w}_{t-1} + \widetilde{\pi}_{t} - \gamma_{w}\widetilde{\pi}_{t-1} + z_{t}^{z}
= \beta z^{1-\sigma} (E_{t}\widetilde{w}_{t+1} - \widetilde{w}_{t} + E_{t}\widetilde{\pi}_{t+1} - \gamma_{w}\widetilde{w}_{t} + E_{t}z_{t+1}^{z})
+ \frac{1 - \xi_{w}}{\xi_{w}} \frac{(1 - \beta \xi_{w}(z^{*})^{1-\sigma})\lambda^{w}}{\lambda^{w} + \chi(1 + \lambda^{w})} \left(\chi \widetilde{\ell}_{t} - \widetilde{\lambda}_{t} - \widetilde{w}_{t} + z_{t}^{b}\right) + z_{t}^{w},$$
(39)

where z_t^w denotes the wage shock, which is defined by

$$z_t^w = \frac{1 - \xi_w}{\xi_w} \times \frac{(1 - \beta \xi_w \bar{z}^{1 - \sigma}) \lambda^w}{\lambda^w + \xi (1 + \lambda^w)} \left(\tilde{\lambda}_t^w + z_t^\ell \right).$$

In the log-linearized equilibrium system, the wage markup shock λ_t^w and the labor supply shocks z_t^ℓ cannot be identified. Then, the wage shock z_t^w is defined as the linear combination of the (log-linearized) wage markup shock $\tilde{\lambda}_t^w$, where λ^w is the steady-state value of λ_t^w and labor supply shock z_t^ℓ .

The monetary policy rule (24) is

$$\tilde{R}_{t}^{n} = \phi_{r} \tilde{R}_{t-1}^{n} + (1 - \phi_{r}) \left\{ \phi_{\pi} \left(\frac{1}{4} \sum_{j=0}^{3} \tilde{\pi}_{t-j} \right) + \phi_{y} (\tilde{y}_{t} - \tilde{y}_{t}^{P}) \right\} + z_{t}^{r}.$$
(40)

The potential output (25) is

$$\tilde{y}_t^P = -\alpha(1+\phi)z_t^z,\tag{41}$$

where $\phi = \Phi/y^*$, and y^* is the steady-state value of detrended output Y_t/Z_t^P . The evolution of aggregate capital (26) is

$$\tilde{k}_{t} = \frac{1-\delta}{z^{*}}(\tilde{k}_{t-1} - z_{t}^{z}) - \frac{\bar{R}^{k}}{z^{*}}\tilde{u}_{t} + \left(1 - \frac{1-\delta}{z^{*}}\right)\tilde{i}_{t}.$$
(42)

The cost-minimization conditions (28) and (29) are

$$\tilde{u}_t + \tilde{k}_{t-1} - \tilde{\ell}_t - z_t^{zp} = \tilde{w}_t - \tilde{R}_t^k,$$
(43)

$$\widetilde{mc}_t = (1 - \alpha)\widetilde{w}_t + \alpha \widetilde{R}_t^k - z_t^{zt}.$$
(44)

The aggregate production function (30) is

$$\tilde{y}_t = (1+\phi) \left\{ z_t^{zt} + (1-\alpha)\tilde{\ell}_t + \alpha(\tilde{u}_t + \tilde{k}_{t-1} - z_t^z) \right\}.$$
(45)

The resource constraint (31) is

$$\tilde{y}_t = \frac{c^*}{y^*} \tilde{c}_t + \frac{i^*}{y^*} \tilde{i}_t + \frac{g^*}{y^*} z_t^g.$$
(46)

There are seven (independent) exogenous shocks in the model. They follow the AR(1)

process:

permanent technology:
$$z_t^{zp} = \rho_{zp} z_{t-1}^{zp} + \sigma_{zp} \varepsilon_t^{zp}$$
 (47)

transitory technology:
$$z_t^{zt} = \rho_{zt} z_{t-1}^{zt} + \sigma_{zt} \varepsilon_t^{zt}$$
 (48)

preference:
$$z_t^b = \rho_b z_{t-1}^b + \sigma_b \varepsilon_t^b$$
 (49)

government purchases:
$$z_t^g = \rho_g z_{t-1}^g + \sigma_g \varepsilon_t^g$$
 (50)

investment adjustment cost:
$$z_t^i = \rho_i z_{t-1}^i + \sigma_i \varepsilon_t^i$$
 (51)

price markup:
$$z_t^p = \rho_p z_{t-1}^p + \sigma_p \varepsilon_t^p$$
 (52)

wage:
$$z_t^w = \rho_w z_{t-1}^w + \sigma_w \varepsilon_t^w$$
 (53)

monetary policy:
$$z_t^r = \rho_r z_{t-1}^r + \sigma_r \varepsilon_t^r$$
. (54)

where $\sigma^x \varepsilon_t^x$ denotes a structural shock to z_t^x for x = zp, zt, b, g, i, p, w, and r, and ε_t^x is independently and identically distributed (i.i.d.), with a mean of zero and standard deviation of one.

3 Data and Estimation Strategy

We use seven quarterly Japanese series as observable variables: real GDP per capita Y_t , real consumption per capita C_t , real investment per capita I_t , real wage W_t , hours worked ℓ_t , consumer price index P_t , and overnight call rate R_t^n . As in Hirose and Kurozumi (2012), the sample period is from 1981:Q1 to 1998:Q4. The model does not take into account the non-linearity of monetary policy. Then, the end of the sample period is set to exclude the zero nominal interest policy of the Bank of Japan.

Except for I_t and ℓ_t , the others series follow Hirose and Kurozumi (2012). I_t is defined as per capita gross fixed capital formation by the private sector, which consists of private residential investment, private non-residential investment, and change in private inventories and is taken from the Economic and Social Research Institute, Cabinet Office of Japan, *the System of National Accounts.* ℓ_t , following Hayashi and Prescott (2002) and Kobayashi and Inaba (2006), is constructed by

$$\ell_t = \frac{\text{Averaged hours worked per employed person} \times \text{Employed person}}{\text{Labor force}}$$

The data source of average hours worked per employed person is the Ministry of Health, Labour and Welfare, *Monthly Labor Survey*. The employed persons and labor force data are from the Ministry of Internal Affairs and Communications, *Labour Force Survey*. Because our study focuses on the labor wedge, this definition of ℓ_t is commonly used for calculating the labor wedge; see, for example, Shimer (2009).

The observation equation is

where $\sigma_j^{me}me_t^j$ denotes the measurement error of variable $j = y, c, i, w, \ell, \pi$, and rn. me_t^j is an i.i.d. shock in which the mean is zero and the standard deviation is one. The measurement errors are necessary to allow the non-existence of structural shocks in the estimation. Generally, the number of exogenous shocks must be equal to or greater than the number of observations. There are seven observations and eight (independent) structural shocks in this model. To allow the non-existence of two or more structural shocks, we need additional exogenous shocks: measurement errors to avoid stochastic singularity.

The growth rate of the labor wedge is calculated by the observations:

$$100\Delta \log \text{Labor Wedge}_t = 100\Delta \log C_t - 100\Delta \log Y_t + (1+\chi)\Delta \log \ell_t.$$
(56)

Most of the model parameters are estimated, while the following aspects are fixed to avoid identification issues. The steady-state depreciation rate of capital stock is set to $\delta(u^*) = 0.06/4$. The cost share of capital in the production function is set to $\alpha = 0.37$. The steady-state wage markup is set to $\lambda^w = 0.20$. These three parameter values are taken from Sugo and Ueda (2008). The steady-state ratio of government purchases to output is set to $g^* = 0.30$, which is at the data mean.

Bayesian estimation is employed by using Dynare. The prior distributions of the parameters are in Tables 1 and 2. For the structural parameters σ , θ , χ , $1/\zeta$, μ , Φ , $\gamma_w, \xi_p, \gamma_p, \xi_p, \phi_r, \phi_{\pi}$, and ϕ_y , the prior distributions are the same as those in Hirose and Kurozumi (2012). For the steady-state price markup λ^p , the mean and standard deviation of the prior distribution are taken from Justiniano, Primiceri, and Tambalotti (2011), but the gamma distribution is employed to fit the theory. For the steady-state growth rate z^* , labor supply ℓ^* , inflation π^* , and real interest r^* , the means of prior distribution are set at the sample mean. For the persistence parameters of structural shocks, the prior distribution is the beta distribution with a mean of 0.5 and a standard deviation of 0.2. For the parameters of standard deviations of structural shocks σ_x and those for measurement errors σ_i^{me} , we consider two types of prior distributions. One is the inverse gamma distribution, which is standard in the literature, with a mean of 0.5 and a standard deviation of infinity. The other is normal distribution. For the parameter of standard deviations of structural shocks σ_x , the mean is 0.1 and the standard deviation is 10. For the parameter of standard deviations of measurement errors σ_j^{me} , the mean is $\sigma_j^{data}/10$ and the standard deviation is σ_j^{data} , where σ_j^{data} is the standard deviation of observations. Because the inverse gamma distribution does not support zero values, it assumes the existence of structural shocks and measurement errors. The normal distribution supports zero value, and then, allows the non-existence of structural shocks. Measurement errors are considered only in the case of normal distribution.

[Tables 1 and 2]

Following standard Bayesian likelihood approaches, the Kalman filter is used to evaluate the likelihood function of the log-linearized equilibrium system and the Metropolis– Hastings algorithm is employed to generate draws from the posterior distribution of the deep parameters. For the ensuing analysis, 1 million draws are generated and the first half of them are discarded. The target of the acceptance rate is about 30%.

4 Empirical Results

The posterior estimates are presented in Tables 3 and 4. Distribution (1) is the case in which inverse gamma distribution is employed as the prior distribution of the parameters for the standard deviation of structural shocks σ_x . Distribution (2) is the case in which normal distribution is employed as the prior distribution of the parameters for the standard deviations of structural shocks σ_x and measurement errors σ_i^{me} .

[Tables 3 and 4]

In Table 3, most estimates are similar in the two prior distributions. Table 4 implies that, in the case of Distribution (2), some structural shocks (transitory technology, preference, government spending, price markup, and wage shocks) might not exist in the sense in which the zero value lies inside the 90% credible intervals of the parameters for the standard deviations. The posterior mean is also close to zero for these parameters except for σ_b (preference shock).

Our result is similar to that of Ferroni, Grassi, and Leon-Ledesma (2019), who find that the posterior means of the parameters for the standard deviations of the government spending, price markup, and wage markup shocks are close to zero in the US economy using Smets and Wouters' (2007) model. Our result is also related to that of Justiniano, Primiceri, and Tambalotti (2013), who find that the variation of the wage markup shock becomes minor if measurement errors are included in the observation equations.⁴

⁴Justiniano, Primiceri, and Tambalotti (2013) employ inverse gamma distribution as a prior distribution

As discussed by Justiniano, Primiceri, and Tambalotti (2013), there is no trade-off between inflation and output stabilization if there is no wage markup shock. Chari, Kehoe, and McGrattan (2009) doubt the existence of a wage markup shock as a fundamental structural shock. Our estimate implies that the wage shock (including wage markup shock) might not exist as their conjecture. However, the labor wedge is driven by other structural shocks in our model.

Table 5 shows the variance decomposition of the output growth $(100\Delta \log Y_t)$ and the labor wedge $(100 \Delta \log \text{ labor wedge})$. As in Tables 3 and 4, Distribution (1) shows the case in which inverse gamma distribution is employed as the prior distribution of standard deviation of structural shocks σ_x , and Distribution (2) is the case in which normal distribution is employed as the prior distribution of standard deviation of structural shocks σ_x and measurement errors σ_i^{me} .

[Table 5]

In the case of Distribution (1), the main driving force of output fluctuation is the investment adjustment cost shock. The preference, government, and permanent technology shocks are also important. The importance of the investment adjustment cost shock is consistent with Justiniano, Primiceri, and Tambalotti (2010, 2011) and Hirose and Kurozumi (2012). However, the main driving force of the labor wedge is the preference and transitory technology shocks. Then, the driving force of the labor wedge is different from that of output fluctuation. According to this result, the investigation of the labor wedge is not promising to understand the business cycle fluctuations.

In the case of Distribution (2), the main driving force of both output fluctuation and the labor wedge is the permanent technology shock. This finding is consistent with Kaihatsu and Kurozumi (2014a, 2014b), who find that the technology shock is the main driving force of business cycles both in the US and Japan. The investment adjustment of the parameters for the standard deviations of shocks, while this study employs normal distribution to allow the non-existence of shocks.

cost shock is also important for these two variables, as in the result of Distribution (1). According to this estimate, output and the labor wedge are mainly driven by the same structural shocks. In this study, Distribution (2) is considered as a better prior distribution, since it allows the non-existence of structural shocks. Therefore, the investigation of the labor wedge fluctuations is promising to understand the business cycles.

Table 6 shows the marginal log likelihood in two estimations: Distributions (1) and (2). According to this table, our prior distribution, the normal distribution, is better than the inverse gamma distribution, which is standard in the literature, as shown by Ferroni, Grassi, and Leon-Ledesma (2019).

5 Concluding remarks

The labor wedge has been investigated by many researchers as an important variable to understand business cycle fluctuations. In this study, we estimated the main sources of the labor wedge and business cycle in the Japanese economy using a canonical mediumscale DSGE model with many nominal and real frictions and many structural shocks. We employed a more flexible prior distribution of the parameters for the standard deviations of structural shocks to allow the non-existence of structural shocks. However, in standard Bayesian estimation, the standard prior distribution of the parameters for the standard deviations of the shocks is inverse gamma distribution, which does not support zero value.

Under the standard prior distribution of the parameters for the standard deviations of structural shocks, our estimation results imply that the labor wedge is mainly driven by preference and transitory technology shocks, whereas the business cycle is mainly driven by the investment adjustment cost shock. Meanwhile, under our relaxed prior distribution, which allows the non-existence of structural shocks, our estimation results show that both the labor wedge and business cycles are mainly driven by the permanent technology and investment adjustment cost shocks. Our results also imply that the investigation of the labor wedge is promising to understand the business cycles, because both the labor wedge and business cycles are driven by the same structural shocks.

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	Parameter	Distribution	Mean	SD
σ	Relative risk aversion	Gamma	1.0000	0.3750
θ	Habit persistence	Beta	0.7000	0.1500
χ	Inverse of Frisch elasticity of labor supply	Gamma	2.0000	0.7500
$1/\zeta$	Inverse of adj. cost of investment	Gamma	4.0000	1.5000
μ	Inverse of elasticity of utilization adj. cost	Gamma	1.0000	1.0000
Φ	Fixed cost in production function	Gamma	0.0750	0.0125
γ_w	Wage indexation	Beta	0.5000	0.2500
ξ_w	Wage stickiness	Beta	0.3750	0.1000
γ_p	Price indexation	Beta	0.5000	0.2500
ξ_p	Price stickiness	Beta	0.3750	0.1000
λ^p	Steady-state price markup	Gamma	0.1500	0.0500
z^*	Steady-state output growth	Gamma	0.353	0.0500
ℓ^*	Steady-state hours worked	Normal	0.0000	0.0500
π^{*}	Steady-state inflation	Gamma	0.341	0.0500
r^*	Steady-state real interest rate	Gamma	1.088	0.0500
ϕ_r	Interest rate smoothing	Beta	0.8000	0.1000
ϕ_{π}	Monetary policy response to inflation	Gamma	1.7000	0.1000
ϕ_y	Monetary policy response to output	Gamma	0.125	0.0500

Table 1: Prior Distribution (1/2)

		Dist	Distribution (1)			Distribution (2)		
	Parameter		Mean	SD		Mean	SD	
$ ho_{zp}$	Persistence of permanent technology shock	Beta	0.5	0.2	Beta	0.5	0.2	
ρ_{zt}	Persistence of transitory technology shock	Beta	0.5	0.2	Beta	0.5	0.2	
$ ho_b$	Persistence of preference shock	Beta	0.5	0.2	Beta	0.5	0.2	
$ ho_i$	Persistence of investment adj. cost shock	Beta	0.5	0.2	Beta	0.5	0.2	
$ ho_{g}$	Persistence of government shock	Beta	0.5	0.2	Beta	0.5	0.2	
$ ho_w$	Persistence of wage shock	Beta	0.5	0.2	Beta	0.5	0.2	
$ ho_p$	Persistence of price markup shock	Beta	0.5	0.2	Beta	0.5	0.2	
$ ho_r$	Persistence of monetary policy shock	Beta	0.5	0.2	Beta	0.5	0.2	
σ_{zp}	SD of permanent technology shock	IG	0.5	Inf	Normal	0.1	10	
σ_{zt}	SD of transitory technology shock	IG	0.5	Inf	Normal	0.1	10	
σ_b	SD of preference shock	IG	0.5	Inf	Normal	0.1	10	
σ_i	SD of investment adj. cost shock	IG	0.5	Inf	Normal	0.1	10	
σ_{g}	SD of government shock	IG	0.5	Inf	Normal	0.1	10	
σ_w	SD of wage shock	IG	0.5	Inf	Normal	0.1	10	
σ_p	SD of price markup shock	IG	0.5	Inf	Normal	0.1	10	
σ_r	SD of monetary policy shock	IG	0.5	Inf	Normal	0.1	10	
σ_y^{me}	SD of ME of GDP	NA	NA	NA	Normal	$\sigma_y^{data}/10$	σ_y^{data}	
σ_c^{me}	SD of ME of consumption	NA	NA	NA	Normal	$\sigma_c^{data}/10$	σ_c^{data}	
σ_i^{me}	SD of ME of investment	NA	NA	NA	Normal	$\sigma_i^{data}/10$	σ_i^{data}	
σ_w^{me}	SD of ME of wage	NA	NA	NA	Normal	$\sigma_w^{data}/10$	σ_w^{data}	
σ^{me}_ℓ	SD of ME of hours worked	NA	NA	NA	Normal	$\sigma_\ell^{data}/10$	σ_ℓ^{data}	
σ_{π}^{me}	SD of ME of inflation	NA	NA	NA	Normal	$\sigma_{\pi}^{data}/10$	σ_{π}^{date}	
σ^{me}_{rn}	SD of ME of call rate	NA	NA	NA	Normal	$\sigma_{rn}^{data}/10$	σ_{rn}^{dat}	

Table 2: Prior Distribution (2/2)

Notes: IG denotes the Inverse Gamma distribution. ME denotes measurement error. σ_x^{data} means the standard deviation of actual data for $x = y, c, i, w, \ell, \pi$, and *rn*.

	Ι	Distribution (1)	Distribution (2)		
	no Measurement Errors		with Measurement Errors		
	(Prior on σ_x : Inverse Gamma)		(Prior on σ_x and σ_j^{me} : Normal)		
	Mean	90% credible interval	Mean	90% credible interval	
σ	1.8176	[0.2985 , 3.3961]	2.3811	[1.6397 , 3.1031]	
θ	0.6795	[0.2730 , 0.9815]	0.3660	[0.1837 , 0.5416]	
X	2.1229	[1.2790 , 3.0523]	1.3495	[0.6600 , 2.0457]	
$1/\zeta$	6.8186	[3.3844 , 10.0621]	4.1703	[1.8334 , 6.3872]	
μ	2.1878	[0.4406 , 3.8635]	0.0705	[0.0000 , 0.1595]	
Φ	0.0776	[0.0559 , 0.0969]	0.0739	[0.0543 , 0.0936]	
γ_w	0.5891	[0.2273 , 0.9791]	0.4347	[0.0248 , 0.8217]	
ξ_w	0.5965	[0.5131 , 0.6786]	0.4664	[0.3246 , 0.6080]	
γ_p	0.4481	[0.1126 , 0.7982]	0.4034	[0.0174 , 0.7739]	
ξ_p	0.705	[0.6407 , 0.7655]	0.5307	[0.3590 , 0.7103]	
λ^p	0.3066	[0.1882 , 0.4234]	0.1494	[0.0687 , 0.2252]	
z^*	0.3222	[0.2530 , 0.3900]	0.4777	[0.3907 , 0.5701]	
ℓ^*	-0.0005	[-0.0828, 0.0820]	-0.0022	[-0.0857, 0.0803]	
π^*	0.3586	[0.2725 , 0.4461]	0.3427	[0.2588 , 0.4236]	
r^*	1.0655	[0.9863 , 1.1400]	1.0119	[0.9367 , 1.0871]	
ϕ_r	0.6897	[0.5943 , 0.7849]	0.6090	[0.5058 , 0.7182]	
ϕ_{π}	1.7645	[1.6089 , 1.9194]	1.7154	[1.5632 , 1.8679]	
ϕ_y	0.1064	[0.0426 , 0.1932]	0.1163	[0.0616 , 0.1697]	

Table 3: Posterior Estimates (1/2)

		Distribution (1)	Ι	Distribution (2)	
	no Measurement Errors		with Measurement Errors		
	(Prior on σ_x : Inverse Gamma)		(Prior on σ_x and σ_j^{me} : Normal)		
	Mean	90% credible interval	Mean	90% credible interval	
$ ho_{zp}$	0.5532	[0.2259 , 0.8721]	0.3854	[0.1779 , 0.5799]	
ρ_{zt}	0.9613	[0.9361 , 0.9864]	0.4739	[0.1031 , 0.8281]	
$ ho_b$	0.3091	[0.0191 , 0.6505]	0.6274	[0.2774 , 0.9948]	
$ ho_i$	0.5005	[0.2382 , 0.7497]	0.8813	[0.8147 , 0.9506]	
$ ho_{g}$	0.7292	[0.3163 , 0.9946]	0.4719	[0.0822 , 0.8315]	
$ ho_w$	0.1049	[0.0138 , 0.1976]	0.7633	[0.3516 , 0.9892]	
$ ho_p$	0.9462	[0.9078 , 0.9853]	0.4963	[0.1661 , 0.8395]	
$ ho_r$	0.539	[0.3479 , 0.7348]	0.4162	[0.2233 , 0.6027]	
σ_{zp}	0.5216	[0.1752 , 0.8636]	1.1108	[0.8277 , 1.5184]	
σ_{zt}	0.8466	[0.7051 , 0.9857]	0.0056	[-0.2459, 0.2548]	
σ_b	9.3614	[0.1110 , 20.7598]	-0.1434	[-1.8223, 1.6020]	
σ_i	5.3242	[3.6466 , 7.0190]	7.1110	[2.9951 , 11.2634]	
σ_{g}	1.5095	[1.0323 , 1.9465]	0.0388	[-0.9004, 0.9858]	
σ_w	0.9992	[0.8245 , 1.1594]	0.0369	[-0.2661, 0.3446]	
σ_p	0.1569	[0.0937 , 0.2253]	-0.0025	[-0.2026, 0.1987]	
σ_r	0.1192	[0.1006 , 0.1368]	0.0939	[0.0714 , 0.1169]	
σ_y^{me}	NA	NA	0.1213	[-0.5871 , 0.6286]	
σ_c^{me}	NA	NA	0.0463	[-0.6649 , 0.6937]	
σ^{me}_i	NA	NA	2.7165	[2.2638 , 3.1748]	
σ_{g}^{me}	NA	NA	0.8827	[0.7491 , 1.0164]	
σ_ℓ^{me}	NA	NA	-0.0972	[-0.4815, 0.4505]	
σ_{π}^{me}	NA	NA	0.0109	[-0.1614, 0.1673]	
σ_{rn}^{me}	NA	NA	0.0007	[-0.0459, 0.0463]	

Table 4: Posterior Estimates (2/2)

Distribution (1)			Distribution (2)		
	no Measurement Errors		with Measurement Errors		
	(Prior on σ^x : Inverse Gamma)		(Prior on σ^x and σ^j_{me} : Normal)		
	$100\Delta \log Y_t$	$100\Delta \log \text{Labor Wedge}_t$	$100\Delta \log Y_t$	$100\Delta \log \text{Labor Wedge}_t$	
z^b	28.02	50.50	0.02	0.38	
z^i	34.93	5.74	12.55	19.59	
z^g	9.68	4.70	0.01	0.08	
z^w	0.76	7.71	0.08	0.95	
z^p	4.99	5.31	0.00	0.00	
z^r	0.21	2.97	0.24	4.53	
z^{zp}	12.92	0.86	85.80	62.40	
z^{zt}	8.49	22.20	0.00	0.05	

Table 5: Variance Decompositions

Table 6: Marginal Log Likelihoods in Two Estimations

Distribution (1)	Distribution (2)		
no Measurement Errors	with Measurement Errors		
(Prior on σ_x : Inverse Gamma)	(Prior on σ_x and σ_j^{me} : Normal)		
-570.393	-549.151		