Mismatch Shocks and Unemployment During the Great Recession

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Abstract

We investigate the macroeconomic consequences of fluctuations in the effectiveness of the labor-market matching process with a focus on the Great Recession. We conduct our analysis in the context of an estimated medium-scale DSGE model with sticky prices and equilibrium search unemployment that features a shock to the matching efficiency (or mismatch shock). We find that this shock is not important for unemployment fluctuations in normal times. However, it plays a somewhat larger role during the Great Recession when it contributes to raise the actual unemployment rate by around 1.3 percentage points and the natural rate by around 2 percentage points. The mismatch shock is the dominant driver of the natural rate of unemployment and explains part of the recent shift of the Beveridge curve.

Keywords: Search and matching frictions; Unemployment; Natural rates.

JEL codes: E32, C51, C52

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1 Introduction

During the Great Recession the unemployment rate in the United States increased markedly from a value of 4.5 percent in mid-2007 to a peak of 10 percent in fall 2009. Since then the labor market has recovered slowly. Nearly three years after its peak, the unemployment rate was still above 8 percent. Some policymakers have related the persistently high rate of unemployment to an increase in sectoral and geographical mismatch between the vacant jobs that are available and the workers who are unemployed (cf. Kocherlakota 2010, among others). This view has received some support from a series of studies that identify a decline in the effectiveness of the process by which the aggregate labor market matched vacant jobs with unemployed workers during the Great Recession (cf. Elsby, Hobijn, and Sahin 2010; Barnichon and Figura 2014, among others). In this paper, we take a general equilibrium perspective and we estimate a medium-scale New Keynesian model with search and matching frictions in the labor market using Bayesian techniques and quarterly data for eight aggregate variables. Our goal is to measure the macroeconomic consequences of the observed decline in matching efficiency – in particular, its impact on the unemployment rate and the unemployment gap.

The spirit of our exercise is quantitative. Our model features the standard frictions and shocks that help in obtaining a good fit of the macro data (Christiano, Eichenbaum, and Evans 2005; Smets and Wouters 2007). In many respects, our model is similar to Gertler, Sala, and Trigari (2008) (henceforth GST) with two main differences: (i) we introduce a shock to the efficiency of the matching function (or "mismatch shock" for short) that we identify by using data on the unemployment rate and the vacancy rate; and (ii) we use the generalized hiring cost function proposed by Yashiv (2000) which combines a pre-match and a post-match component. We discuss the two deviations from GST (2008) in turn.

Matching efficiency shocks are already present in the seminal paper by Andolfatto (1996), which interprets them as sectoral reallocation shocks of the kind emphasized by Lilien (1982). These shocks can be seen as the Solow residual of the matching function and as catch-all shocks for structural changes in the labor market such as the degree of skill mismatch between jobs and workers (Sahin, Song, Topa, and Violante 2014; Herz and van Rens 2011); the importance of geographical mismatch that might have been exacerbated by house-locking effects (Sterk 2011); workers’ search intensity that may have been reduced by the extended duration of unemployment.

\footnote{More recently, DSGE models featuring matching efficiency shocks have been considered by Beauchemin and Tasci (2014), Cheremukhin and Restrepo-Echavarria (2014), Justiniano and Michelacci (2011), Krause, Lopez-Salido, and Lubik (2008) and Lubik (2013) among others.}
benefits (Fujita 2011; Nakajima 2012; Zhang 2013); firms’ recruiting efforts (Davis, Faberman, and Haltiwanger 2013); and shifts in the composition of the unemployment pool, such as a rise in the share of long-term unemployed, or fluctuations in participation due to demographic factors (Barnichon and Figura 2014). If these structural factors played an important role during the Great Recession, matching efficiency shocks should emerge as a prominent driver of the surge in the unemployment rate. Our goal is to quantify their contribution.

In our model, firms’ hiring costs consist of a pre-match and a post-match component. The pre-match component is the search cost of advertising vacancies, a standard ingredient of models with search and matching frictions in the labor market (Pissarides 2000). The post-match component is the cost of adjusting the hiring rate. We can think of it as capturing training costs (GST 2008). We combine the two hiring costs components because the nature of hiring costs is crucial for the propagation of matching efficiency shocks. In particular, when firms do not face any pre-match costs, as in GST (2008), matching shocks exert no effects on the unemployment rate. In contrast, the unemployment rate fluctuates significantly in response to matching efficiency shocks when firms face pre-match hiring costs only. Therefore, the share of pre-match costs in total hiring costs is a key parameter that governs the propagation of matching efficiency shocks and that we estimate in our analysis.

We find that matching efficiency shocks do not play an important role for business cycle fluctuations. They generate a positive conditional correlation between unemployment and vacancies while the two variables are strongly negatively correlated in the data. Nevertheless, these shocks play a somewhat larger role during and after the Great Recession when matching efficiency declines substantially and unemployment and vacancies move in the same directions for few quarters. In this episode mismatch shocks explain about 1.3 percentage points of the increase in the unemployment rate, a result that is in the ballpark of the values found by studies using disaggregated data (cf. Barnichon and Figura 2014, Sahin, Song, Topa, and Violante 2014). Our results suggest that the bulk of the rise in the unemployment rate during the Great Recession is driven by a series of negative demand shocks, in particular risk premium shocks and investment-specific shocks. Nevertheless, negative matching efficiency shocks contribute to weaken the recovery in the aftermath of the Great Recession and to explain the shift in the Beveridge curve.

From the perspective of a monetary policymaker, looking at the drivers of the actual unemployment rate is not sufficient. As Kocherlakota (2011) puts it, monetary policy should focus on
offsetting the effects of nominal rigidities. To do so, monetary policy may aim at closing the gap between the actual and the natural rate of unemployment. As stressed by Kocherlakota (2011), a big challenge for policymakers is that the natural rate is unobservable and fluctuates over time. To address this issue, we use our estimated model to infer the path of the natural rate. We define the natural rate as the counterfactual rate of unemployment that emerges in a version of the model with flexible prices and wages, constant price mark-up, and constant bargaining power, in keeping with the previous literature (Smets and Wouters 2007; Sala, Söderström and Trigari 2008; Groshenny 2013). Even though matching efficiency shocks have limited importance for fluctuations in actual unemployment, we find that these shocks are a dominant source of variation in the natural rate. This result is due to the fact that nominal rigidities dampen the propagation of matching efficiency shocks and enhance the effects of all the other shocks. We find that the deterioration in the effectiveness of the aggregate labor market matching process during the Great Recession contributes to raising the natural rate by about 2 percentage points. Hence, negative matching efficiency shocks help close the gap between the actual and the natural rate of unemployment. The model indicates that in 2013:Q2 the natural and the actual rate almost coincide slightly below 8 percent.

Our paper is related to two strands of the literature. We contribute to the literature initiated by Lilien (1982) on the importance of reallocation shocks as a source of unemployment fluctuations. Abraham and Katz (1986) and Blanchard and Diamond (1989) look at shifts in the sectoral composition of demand and estimate a series of regressions to disentangle the importance of reallocation shocks and aggregate demand shocks. Both papers emphasize the primacy of aggregate demand shocks in producing unemployment fluctuations and find that reallocation shocks are almost irrelevant at business cycle frequencies. Our contribution to this literature is the use of an estimated dynamic stochastic general equilibrium model (DSGE), rather than a reduced-form model, with a focus on the role of the nominal rigidities and the hiring cost function.

Our paper also relates to the literature that studies the output gap derived from estimated New Keynesian models (Smets and Wouters 2007; Justiniano, Primiceri, and Tambalotti 2013). Often in this literature, the labor market is modeled only along the intensive margin (hours worked). Notable exceptions are Galí, Smets and Wouters (2011) and Sala, Söderström, and Trigari (2008). Galí, Smets, and Wouters (2011) estimate a model with unemployment and also compute a measure of the natural rate. However, in their model, unemployment is due only to
the presence of sticky wages (there are no search and matching frictions) so that the natural rate fluctuates only in response to wage mark-up shocks. In our model, unemployment is due to both nominal rigidities and search and matching frictions and wage mark-up shocks play a limited role. Moreover, our measure of the natural rate fluctuates in response to several shocks. Sala, Söderström, and Trigari (2008) provide a similar model-based measure of the natural rate. Their model, however, does not feature matching efficiency shocks which are, according to our estimates, prominent drivers of the natural rate.

The paper proceeds as follows: Section 2 lays out the model. Section 3 explains our econometric strategy. Section 4 discusses the transmission mechanism of mismatch shocks. Section 5 presents our empirical results. Section 6 relates to the sensitivity analysis. Finally, Section 7 concludes.

2 Model

Our model builds upon GST (2008) and Groshenny (2013) and merges the New Keynesian model with the search and matching model of unemployment. The model incorporates the standard features introduced by Christiano, Eichenbaum, and Evans (2005) to help the model fit the postwar U.S. macro data. Moreover, as in the benchmark quantitative macroeconometric model of Smets and Wouters (2007), fluctuations are driven by multiple exogenous stochastic disturbances. GST (2008) have shown that such a model fits the macro data as accurately as the Smets and Wouters (2007) model.

In addition to the inclusion of the matching efficiency shock and the use of the generalized hiring cost function, our model features other small differences compared to GST (2008). First, as in Smets and Wouters (2007), we have a risk premium shock, rather than a preference shock, to capture disturbances originating in the financial markets. Given that financial disruptions are a defining feature of the Great Recession, we believe it is important to have a financial shock in the model. Second, we use the timing proposed by Ravenna and Walsh (2008) in the law of motion for employment: new hires become productive in the current period and separated workers start searching for a job immediately so that they do not have necessarily to be unemployed for one period. This specification implies that employment is not a pre-determined variable and delivers higher unemployment volatility. Third, we simplify the model in some dimensions that are not essential for our analysis by using quadratic adjustment costs in prices and wages instead of staggered time-dependent contracts. We also use a Dixit-Stiglitz
aggregator with constant elasticity of substitution across goods instead of the Kimball aggregator with endogenous elasticity.

**The representative household.** There is a continuum of identical households of mass one. Each household is a large family, made up of a continuum of individuals of measure one. Family members are either working or searching for a job. We assume that family members pool their income before allowing the head of the family to optimally choose per capita consumption.

The representative family enters each period $t = 0, 1, 2, \ldots$, with $B_{t-1}$ bonds and $K_{t-1}$ units of physical capital. Bonds mature at the beginning of each period, providing $B_{t-1}$ units of money. The representative family uses some of this money to purchase $B_t$ new bonds at nominal cost $B_t/R_t$, where $R_t$ denotes the gross nominal interest rate.

The representative household owns the stock of physical capital $K_t$ which evolves according to

$$K_t \leq (1 - \delta) K_{t-1} + \mu_t \left[ 1 - L \left( \frac{I_t}{I_{t-1}} \right) \right] I_t,$$

where $\delta$ denotes the depreciation rate. The function $L$ captures the presence of adjustment costs in investment. An investment-specific technology shock $\mu_t$ affects the efficiency with which consumption goods are transformed into capital. The shock follows an autoregressive process of order one as all the other seven shocks in the model.

The household chooses the capital utilization rate, $u_t$, which transforms physical capital into effective capital according to $K_t = u_t K_{t-1}$. The household faces a cost $a(u_t)$ of adjusting the capacity-utilization rate. The household rents effective capital services to firms at the nominal rate $r^K_t$.

Each period, $N_t$ family members are employed. Each employee works a fixed amount of hours and earns the nominal wage $W_t$. The remaining $(1 - N_t)$ family members are unemployed and each receives nominal unemployment benefits $b_t$, financed through lump-sum taxes. Unemployment benefits $b_t$ are proportional to the nominal wage along the steady-state balanced growth path $b_t = \tau W_{ss,t}$. The fact that unemployment benefits grow along the balanced growth path ensures that unemployment remains stationary. During period $t$, the representative household receives total nominal factor payments $r^K_t K_t + W_t N_t + (1 - N_t) b_t$ as well as profits $D_t$. The family uses these resources to purchase finished goods for both consumption and investment purposes.
The family’s period $t$ budget constraint is given by

$$P_tC_t + Pt_I + \frac{B_t}{\epsilon_{bt}R_t} \leq B_{t-1} + W_tN_t + (1 - N_t)b_t + r_t^K u_t K_{t-1}$$

$$-Pt a(u_t) K_{t-1} - T_t + D_t. \tag{2}$$

As in Smets and Wouters (2007), the shock $\epsilon_{bt}$ drives a wedge between the central bank’s instrument rate $R_t$ and the return on assets held by the representative family.

The family’s lifetime utility is described by

$$E_t \sum_{s=0}^{\infty} \beta^s \ln (C_{t+s} - hC_{t+s-1}), \tag{3}$$

where $0 < \beta < 1$ and $h > 0$ captures internal habit formation in consumption.

**The representative intermediate goods-producing firm.** Each intermediate goods-producing firm $i \in [0, 1]$ enters in period $t$ with a stock of $N_{t-1}(i)$ employees. Before production starts, $\rho N_{t-1}(i)$ old jobs are destroyed. The job destruction rate $\rho$ is constant. Those workers who have lost their jobs start searching immediately and can potentially still be hired in period $t$ (Ravenna and Walsh 2008). Employment at firm $i$ evolves according to $N_t(i) = (1 - \rho) N_{t-1}(i) + m_t(i)$, where the flow of new hires $m_t(i)$ is given by $m_t(i) = q_t \nu_t(i)$. $\nu_t(i)$ denotes vacancies posted by firm $i$ in period $t$ and $q_t$ is the aggregate probability of filling a vacancy, $q_t = \frac{m_t}{\nu_t}$, where $m_t = \int_0^1 m_t(i) \, di$ and $\nu_t = \int_0^1 \nu_t(i) \, di$ denote aggregate matches and vacancies respectively. Aggregate employment $N_t = \int_0^1 N_t(i) \, di$ evolves according to

$$N_t = (1 - \rho) N_{t-1} + m_t. \tag{4}$$

The matching process is described by an aggregate constant-returns-to-scale Cobb-Douglas matching function,

$$m_t = \zeta_t S_t^\sigma \nu_t^{1-\sigma}, \tag{5}$$

where $S_t$ denotes the pool of job seekers in period $t$, $S_t = 1 - (1 - \rho) N_{t-1}$, and $\zeta_t$ is a time-varying scale parameter that captures the efficiency of the matching technology. It evolves exogenously following the autoregressive process,

$$\ln \zeta_t = (1 - \rho_\zeta) \ln \zeta + \rho_\zeta \ln \zeta_{t-1} + \epsilon_{\zeta t}. \tag{6}$$
where $\zeta$ denotes the steady-state matching efficiency and $\varepsilon_{\zeta t} \sim i.i.d. N\left(0, \sigma_{\zeta}^2\right)$. Aggregate unemployment is defined by $U_t \equiv 1 - N_t$.

Firms face hiring costs $H_t(i)$ measured in terms of the finished good and given by a generalized hiring function proposed by Yashiv (2000) that combines a pre-match and a post-match component in the following way,

$$H_t(i) = \frac{\kappa}{2} \left( \phi_V V_t(i) + (1 - \phi_V) m_t(i) \right)^2 Y_t,$$

(7)

where $\kappa$ determines the output-share of hiring costs and $0 \leq \phi_V \leq 1$ governs the relative importance of the pre-match component. When $\phi_V$ is equal to 0 we are back to the model with only post-match hiring costs (GST 2008). Instead, when $\phi_V$ is equal to 1 we obtain a model with quadratic pre-match hiring costs (Pissarides 2000). Interestingly, the empirical literature has so far preferred a specification with post-match hiring costs, that can be interpreted as training costs.

Each period, firm $i$ combines $N_t(i)$ homogeneous employees with $K_t(i)$ units of efficient capital to produce $Y_t(i)$ units of intermediate good $i$ according to the constant-returns-to-scale technology described by

$$Y_t(i) = A_t^{1-\alpha} K_t(i)^{\alpha} N_t(i)^{1-\alpha}.$$

(8)

$A_t$ is an aggregate labor-augmenting technology shock whose growth rate, $z_t \equiv A_t/A_{t-1}$, follows an exogenous process.

Intermediate goods substitute imperfectly for one another in the production function of the representative finished goods-producing firm. Hence, each intermediate goods-producing firm $i \in [0, 1]$ sells its output $Y_t(i)$ in a monopolistically competitive market, setting $P_t(i)$, the price of its own product, with the commitment of satisfying the demand for good $i$ at that price. Each intermediate goods-producing firm faces costs of adjusting its nominal price between periods, measured in terms of the finished good and given by

$$\frac{\phi_P}{2} \left( \frac{P_t(i)}{\pi_{t-1}^{1-\epsilon} P_{t-1}(i)} - 1 \right)^2 Y_t.$$

(9)

The term $\phi_P$ governs the magnitude of the price adjustment cost. The expression $\pi_t = \frac{P_t}{P_{t-1}}$ denotes the gross rate of inflation in period $t$. The steady-state gross rate of inflation is denoted
by $\pi > 1$ and coincides with the central bank’s target. The parameter $0 \leq \zeta \leq 1$ governs the importance of backward-looking behavior in price setting (Ireland 2007).

We model nominal wage rigidities as in Arsenau and Chugh (2008). Each firm faces quadratic wage-adjustment costs which are proportional to the size of its workforce and measured in terms of the finished good,

$$\frac{\phi W}{2} \left( \frac{W_t (i)}{z \pi_{t-1}^{\theta} \pi^{1-\theta} W_{t-1} (i)} - 1 \right)^2 N_t (i) Y_t, \tag{10}$$

where $\phi W$ governs the magnitude of the wage adjustment cost. The parameter $0 \leq \vartheta \leq 1$ governs the importance of backward-looking behavior in wage setting. Firms take the nominal wage as given when maximizing the discounted value of expected future profits.

**Wage setting.** The nominal wage $W_t (i)$ is determined through surplus sharing,

$$W_t (i) = \arg \max \left( \Delta_t (i)^{\eta_t} J_t (i)^{1-\eta_t} \right). \tag{11}$$

The worker’s surplus, expressed in terms of final consumption goods, is given by

$$\Delta_t (i) = \frac{W_t (i)}{P_t} - \frac{b_t}{P_t} + \beta \chi E_t (1 - s_{t+1}) \frac{\Lambda_{t+1}}{\Lambda_t} \Delta_{t+1} (i), \tag{12}$$

where $\chi = 1 - \rho$. $\Lambda_t$ denotes the household’s marginal utility of wealth and $s_t = m_t / S_t$ is the aggregate job finding rate. The firm’s surplus in real terms is given by

$$J_t (i) = \xi_t (i) (1 - \alpha) \frac{Y_t (i)}{N_t (i)} - \frac{W_t (i)}{P_t} - \frac{\phi W}{2} \left( \frac{W_t (i)}{z \pi_{t-1}^{\theta} \pi^{1-\theta} W_{t-1} (i)} - 1 \right)^2 Y_t + \kappa \frac{\Lambda_{t+1}}{\Lambda_t} \Xi_t (i), \tag{13}$$

where $\xi_t (i)$ denotes firm $i$’s real marginal cost. The worker’s bargaining power $\eta_t$ evolves exogenously and $0 < \eta < 1$ denotes the steady-state worker’s bargaining power.

**The representative finished goods-producing firm.** During each period $t = 0, 1, 2, \ldots$, the representative finished good-producing firm uses $Y_t (i)$ units of each intermediate good $i \in [0, 1]$, purchased at the nominal price $P_t (i)$, to produce $Y_t$ units of the finished good according to the constant-returns-to-scale technology described by

$$\left( \int_0^1 Y_t (i)^{\theta_t (i)/\theta_t - 1} dt \right)^{\theta_t (i)/\theta_t - 1} \geq Y_t, \tag{14}$$
where $\theta > 1$ is the demand elasticity and $\theta_t$ is an exogenous process for the demand elasticity that translates in exogenous variations in the price markup.

**Monetary and fiscal authorities.** The central bank adjusts the short-term nominal gross interest rate $R_t$ by following a Taylor-type rule similar to the one proposed by Justiniano, Primiceri and Tambalotti (2013):

$$
\ln \frac{R_t}{\bar{R}} = \rho_r \ln \frac{R_{t-1}}{\bar{R}} + (1 - \rho_r) \left( \rho_\pi \ln \frac{(P_t/P_{t-4})^{1/4}}{\pi} + \rho_y \ln \frac{(Y_t/Y_{t-4})^{1/4}}{z} \right) + \ln \epsilon_{mpt}.
$$

(15)

The degree of interest-rate smoothing $\rho_r$ and the reaction coefficients $\rho_\pi$ and $\rho_y$ are all positive. The monetary policy shock $\epsilon_{mpt}$ follows an exogenous process.

The government budget constraint takes the form,

$$
P_t G_t + (1 - N_t) b_t = \left( \frac{B_t}{R_t} - B_{t-1} \right) + T_t,
$$

(16)

where $T_t$ denotes total nominal lump-sum transfers. Public spending is an exogenous time-varying fraction of GDP, $G_t = \left( 1 - \frac{1}{\epsilon_{gt}} \right) Y_t$, where $\epsilon_{gt}$ follows an exogenous process.

Details on the first order conditions, the log-linearized system, the calibration, solution and estimation of the model are provided in the online appendix.

### 3 Econometric Strategy

We calibrate 13 parameters that we report in Table 1. Since these values follow the previous literature, we focus here only on the parameters related to the labor market. The vacancy-filling rate is set equal to 0.70, which is just a normalization. Calibrated values for the steady state quarterly separation rate range in the literature from 0.05 in Krause, López-Salido, and Lubik (2008) to 0.15 in Andolfatto (1996). We use the conventional value 0.085, in line with most of the literature (Yashiv 2006). We set the elasticity of the matching function with respect to unemployment at 0.65 in the middle of the range of values estimated in five recent studies (Barnichon and Figura 2014; Justiniano and Michelacci 2011; Lubik 2013; Shimer 2005; Sedlacek 2014). The calibration of the replacement rate at 0.4 is a conservative choice based on Shimer (2005) and Yashiv (2006).

We estimate the remaining 27 parameters using Bayesian techniques. Our priors, summarized in Tables 2 and 3, are standard (Smets and Wouters 2007; GST 2008). The estimation period is 1957:Q1–2008:Q3. Following Galí, Smets and Wouters (2011) we stop our sample period in
the beginning of the Great Recession to prevent our estimates from being distorted by the zero-lower bound and by other non-linearities. Nevertheless, we rely on the estimated parameters to simulate the model using data until 2013:Q2 to discuss the behavior of the aggregate variables in the recent turbulent years; that is, beyond the sample period.

The model includes as many shocks as observables. The estimation uses quarterly data on eight key macro variables. We downloaded seven series from the FREDII database maintained by the Federal Reserve Bank of St. Louis. We measure nominal consumption using data on nominal personal consumption expenditures of nondurables and services. Nominal investment corresponds to the sum of personal consumption expenditures of durables and gross private domestic investment. Nominal output is measured by nominal GDP. Per capita real GDP, consumption, and investment are obtained by dividing the nominal series by the GDP deflator and population. Real wages correspond to nominal compensation per hour in the nonfarm business sector, divided by the GDP deflator. Consistently with the model, we measure population by the labor force which is the sum of official unemployment and official employment. The unemployment rate is the official unemployment divided by the labor force. Inflation is the first difference of the log of the GDP deflator. The nominal interest rate is measured by the effective federal funds rate.

Our eighth observable variable is the vacancy rate. As in Justiniano and Michelacci (2011), data on job vacancies are used to construct the vacancy rate as the ratio of job vacancies over the sum of job vacancies and employment, consistent with the definition of job opening rate used in JOLTS. The series for job vacancies is taken from Barnichon (2010) who constructs (and updates regularly) a new vacancy index that combines the print Help-Wanted Index with the online Help-Wanted index published by the Conference Board since 2005. The series tracks closely the (rescaled) JOLTS measure of job openings that starts in December 2000. As emphasized by Shimer (2005), a shortcoming of the print Help-Wanted index is that it is subject to low frequency fluctuations that are related only tangentially to the labor market. On the one hand, the Internet may have reduced the reliance of firms on using advertising in newspapers well before 2005. On the other hand, Shimer (2005) describes how a series of newspaper consolidations and Equal Opportunity laws may have encouraged firms to rely more extensively on newspaper advertising in the first part of the sample. Therefore, to remove these secular shifts we follow Shimer (2005), Justiniano and Michelacci (2011), Davis, Faberman and Haltiwanger (2013) and we detrend the vacancy rate series using an HP filter with smoothing weight equal to $10^{-6}$.
In Tables 2 and 3 we report the outcome of our estimation exercise. Since most estimates are in line with the previous literature, we concentrate our attention on the parameters related to the labor market. The weight of the pre-match component in the convex combination $\phi_V$ is estimated at 0.35 at the posterior median. Although we use an agnostic prior centered around 0.5, the data favor a large post-match component, as it has been found in the previous literature. In fact, Yashiv (2000) estimates a value of 0.3 on Israeli data using the same functional form. Sala, Söderström and Trigari (2013) and Christiano, Trabandt and Walentin (2011) use a different functional form and find a slightly lower weight on the pre-match component using US and Swedish data respectively. The posterior mode of steady state hiring costs as a percent of output is estimated at 0.25 percent. This corresponds to 4.5 percent of total wages of newly hired workers, thus inside the range between 4 and 14 percent documented by Silva and Toledo (2009). Christiano, Eichenbaum and Trabandt (2014) estimate a value of 0.5 percent.

Our model provides a rather conventional view on business cycle fluctuations over the sample period, as reported in the infinite horizon variance decomposition in Table 4. The relevant sources of output fluctuations in the model are neutral technology shocks, investment-specific technology shocks, and risk-premium shocks. Our results are consistent with GST (2008) once we take into account that the risk premium shock limits somewhat the importance of the investment specific technology shock. A less conventional implication of our model is that wage-bargaining shocks do not matter for output fluctuations. This result was already present in GST (2008) but, as far as we know, it has not been commented in the literature. Chari, Kehoe, and McGrattan (2009) have criticized the New Keynesian model for its reliance on dubiously structural shocks such as the wage-bargaining (or wage mark-up) shock. Here, we find that this criticism does not apply. Our finding suggests that search and matching frictions in the labor market, and the use of labor market variables in the estimation, absorb the explanatory power of the wage-bargaining shock. Put differently, our estimated DSGE model seems successful at endogenizing the labor wedge.

4 Inspecting the Mechanisms: the Role of Hiring Costs and Nominal Rigidities

In this section we concentrate on the macroeconomic effects of matching efficiency shocks. In Figure 1 we plot impulse responses computed at the posterior mode (bold lines). When matching
efficiency declines, the probability of filling a vacancy drops and hiring becomes more expensive since more vacancies have to be posted to hire a worker. In response to the increase in costs firms hire fewer workers and, given the assumption of instantaneous hiring, employment and output decline already on the impact of the shock while unemployment increases. A lower probability of filling a vacancy increases the hiring cost component of the marginal cost and partly translates into higher prices through the New Keynesian Phillips curve. Vacancies increase in our model and thus positively comove with unemployment in response to mismatch shocks. Why do firms post more vacancies despite hiring fewer new workers? Essentially for two reasons: i) because posting vacancies is relatively inexpensive and ii) because the decline in hiring is limited, mitigated by the presence of nominal rigidities. In our model posting vacancies is relatively inexpensive because pre-match hiring costs account for a limited share of total hiring costs. The lower this share is, the lower is the propagation of the shock: in Figure 1 we show this point by proposing two counterfactuals in which we change only the value of the parameter $\phi_V$ and we maintain all the other parameter values at their posterior mode. The dashed-solid line refers to a model with almost only post-match hiring costs ($\phi_V = 0.01$): in this case posting vacancies is almost costless and firms react by posting many more vacancies than in our baseline case so that they are able to undo the effects of the shock and output and unemployment are barely affected. This extreme case corresponds to the GST (2008) model and shows that output and unemployment are invariant to mismatch shocks in that set-up. In other words, in the absence of a pre-match component a decline in matching efficiency has no effects on the macroeconomy. The other polar case is with almost only pre match hiring costs ($\phi_V = 0.99$, dashed lines): in this case posting vacancies is relatively expensive, vacancies increase less than in our baseline case and the shock has a more contractionary effect on hiring and on output.

The second reason that favors a positive response of vacancies is the presence of nominal rigidities (the estimated average duration of prices in our model is 4 quarters when using the metric of the Calvo model). In Figure 2 we compare our estimated model (solid lines) with a counterfactual where we impose flexible prices and wages and we maintain all the other parameters at their estimated value (dashed-solid line). When prices are flexible, firms can increase prices optimally to restore their profits impaired by the increase in costs, so as to keep markups constant. With higher prices the fall in aggregate demand is more pronounced and firms need a larger contraction in hiring. This leads to a larger response of unemployment and to a lower increase in vacancies than in the baseline case with nominal rigidities.\textsuperscript{2}

\textsuperscript{2}When combining the two effects (high share of pre-match hiring costs and low degree of nominal rigidities)
5 Empirical Results

In this section we use our model to investigate the importance of mismatch shocks over the sample period and, more specifically, beyond the sample period during and after the Great Recession. We then provide a model based estimate of the natural rate of unemployment.

5.1 Mismatch Shocks, Unemployment and Vacancies

The estimated series for mismatch shocks is plotted in Figure 3: it reaches its minimum in the beginning of the 80s and then it starts improving around 1985 until 2002 when it peaks. The improvement in matching efficiency could reflect the firms more widely adopting information technologies (the so-called New Economy). After 2002 matching efficiency declines, with a substantial acceleration during the Great Recession, and stays at unprecedentedly low levels in recent years. These dynamics are broadly consistent with Barnichon and Figura (2014) who estimate matching efficiency by regressing the job finding rate over the labor market tightness, although they identify an even larger decline in the recent years.

Matching efficiency shocks explain only 6 percent of unemployment volatility (cf. Table 4) whereas they are almost irrelevant for output fluctuations over the sample period (1957:Q1-2008:Q3). This is not so surprising since mismatch shocks generate a large positive correlation between unemployment and vacancies whereas in the data the two series are strongly negatively correlated. A limited importance of matching efficiency shocks is consistent with fact that the data favor a substantial degree of nominal rigidities and a limited share of pre-match hiring costs and, as we have discussed in the previous section, both features tend to dampen the propagation of the shocks. Moreover, the Great Recession, a period of large fluctuations in matching efficiency, is not part of the sample period for estimation, thus maintaining low the share of variance explained by mismatch shocks.

The limited importance of mismatch shocks for business cycle fluctuations in general does not rule out that these shocks may play a relevant role in specific episodes, in particular when unemployment and vacancies move in the same direction, as in the aftermath of the Great Recession. We now make use of our estimated model to discuss the dynamics of aggregate variables over the period 2008:Q4-2013:Q2. In Figure 4 we plot the historical decomposition of the unemployment rate. Since 2009, negative mismatch shocks are responsible on average for vacancies may even decline in response to a negative matching efficiency shock, thus leading to a relatively large decline in output. In that scenario, that however is in contrast with our parameter estimates, mismatch shocks may generate a negative conditional correlation between unemployment and vacancies.
about 1.3 percentage points of the large increase in the unemployment rate.\footnote{Sala, Söderstöm, and Trigari (2013) conduct the same experiment in a similar model with a focus on a cross-country comparison. They find results that are in line with ours for the United States.} The contribution of these shocks is limited in the most acute phase of the crisis but is more relevant in the slow phase of recovery. This result is in line with studies using disaggregated data. Barnichon and Figura (2014) who decompose movements in the Beveridge curve and conclude that without any loss in matching efficiency, unemployment would have been about 150 points lower in late 2010. Sedlacek (2014) finds results very similar to ours. Sahin, Song, Topa, and Violante (2014) confine their attention to the more narrow concept of mismatch unemployment. They combine disaggregated data to construct a mismatch index and they find that mismatch unemployment at the 2-digit industry level can account for 0.75 percentage points out of the 5.4 increase in the U.S. unemployment rate from 2006 to the Fall 2009. This result is compatible with our evidence, given that mismatch is not the only driver of matching efficiency.

From Figure 4 we see that the large increase in unemployment during the Great Recession is explained by a series of large negative demand shocks like risk-premium shocks (in particular during 2009) and investment shocks. Fiscal policy shocks have contributed materially to lower unemployment, reflecting the effects of the fiscal stimulus package implemented by the U.S. authorities in the aftermath of the crisis and perhaps also some foreign shocks. Finally, we find that negative bargaining power shocks (that is, a reduction in the bargaining power of workers) have contributed to lowering the unemployment rate throughout the Great Recession (and during the entire previous decade). This finding may reflect competitive pressures from abroad and threats of offshoring from the domestic market. Arsenau and Leduc (2012) show how the threat to offshore can have large effects on wages even when the actual amount of offshoring in the economy is small.

An alternative way to evaluate the role of mismatch shocks during the Great Recession is to consider the Beveridge curve. In Figure 5 the grey dots describe the joint evolution of unemployment and vacancies in the data over the period 2008:Q1-2013:Q2. The variables are expressed in percentage deviation from their value in 2008:Q1, when the unemployment rate was at 5 percent. The black solid line connects the counterfactual values for unemployment and vacancies obtained from our model when we turn off only matching efficiency shocks. We see that the combination of the remaining seven shocks can replicate the shape of the Beveridge curve, thus showing that standard shocks can generate shifts in the Beveridge curve by themselves. However, matching efficiency shocks are essential to match the shift from a quantitative point of
view, in particular towards the end of the sample when the gap between the two loops widens. In particular, in the last quarter of our sample (2013:Q2) unemployment was around 7.5 (more than 40 percent higher than its value in 2008:Q1). The counterfactual in absence of matching efficiency shocks would predict a value of around 5.5 percent (only 10 percent higher than its value in 2008:Q1). These results are interesting in light of a recent paper by Christiano, Eichenbaum and Trabandt (2014) who show that a model without matching efficiency shocks can generate a shift in the Beveridge curve and explain the data. Here we confirm their result but we show also that, once matching efficiency shocks are introduced, they are important to explain unemployment dynamics during the Great Recession and even more in its aftermath, unlike in other periods when they are often irrelevant. Lubik (2013) finds results similar to ours in a Real Business Cycle model.

5.2 Mismatch Shocks and the Natural Rate

The natural rate of unemployment is unobservable and its estimation is a main challenge for monetary policymakers. In this section, we use our estimated medium-scale DSGE model to infer the path of the natural rate and, unlike in the previous literature, we discuss the role of mismatch shocks in its dynamics. Following Sala, Söderström, and Trigari (2008), Groshenny (2013) and the related literature on the output gap in DSGE models, we define the natural rate to be the unemployment rate that would prevail if i) prices and wages were perfectly flexible and ii) the markup of price over marginal cost and the bargaining power of workers were constant.

We adopt the standard practice of turning off the inefficient shocks to compute the natural rate. Price mark-up shocks and bargaining power shocks are inefficient. The former ones affect the degree of imperfect competition in the goods market. The latter shocks induce deviations from the Hosios condition and so affect the severity of the congestion externality that characterizes the labor market in the search and matching model. This standard definition is in line with the concept of natural rate expressed in Friedman (1968), i.e. a measure of unemployment that fluctuates over time in response to shocks and that is independent from monetary factors. Moreover this definition is also shared by some monetary policymakers. For example, it is consistent with Kocherlakota (2011)’s view of the Fed’s mission.¹²

¹²The shocks affecting the natural rate are technology, investment-specific, fiscal and matching efficiency shocks. Monetary and risk premium shocks leave the natural rate unaffected because they do not propagate under flexible prices and wages.

¹³Our approach is common in the literature but is not uncontroversial. In particular, the interpretation of labor supply shocks in the New Keynesian model is the object of a recent literature (Chari, Kehoe, and McGrattan 2009; Gali, Smets, and Wouters 2011; Justiniano, Primiceri, and Tambalotti 2013) but is outside the scope of
In Figure 6 we plot the observed unemployment rate together with our estimate of the natural rate. If we focus on the very low frequencies just for a while, we see that the natural rate was gently trending upward until 1980, and then had been gradually decreasing, reaching a trough around 2003. Our estimate of the natural rate is rather precisely estimated, in contrast with Staiger, Stock, and Watson (1997) who argue that large confidence bands are a distinguishing feature of the natural rate. Not so surprisingly, we find that the cross-equation restrictions embedded in our estimated DSGE model provide quite a sharp identification strategy of the unobserved natural rate.

Interestingly, according to our model actual unemployment was well below the natural rate over the period 2005–2007, thus signaling some overheating in the economy. During the Great Recession the posterior median estimate of the natural rate rises sharply but reaches its peak as late as in the beginning of 2013. Therefore, while actual unemployment has been declining since 2010, the natural rate has been increasing until the end of our sample. How can we rationalize this diverging behavior between the actual and the natural rate? The answer is in Figure 7 where we see that large negative matching efficiency shocks (that have a more limited impact on the actual rate as shown in Figure 4) lead to an increase in the natural rate of almost 2 percentage points since the beginning of the Great Recession with larger effects in recent years.

More generally, we remark that mismatch shocks are the dominant source of variation in the natural rate. Why are mismatch shocks so important for the natural rate when they have a more limited effect on the actual rate? Simply because, as already highlighted in Figure 2, mismatch shocks propagate more under flexible prices and wages. The dashed-solid line in fourth panel of Figure 2 is in fact the impulse response of the natural rate of unemployment. While the natural rate reacts more than the actual rate to mismatch shocks, the opposite is true for the other shocks (neutral technology, investment-specific, and government spending shocks) that propagate little under flexible prices and wages, as shown by Shimer (2005) and in the following literature on the so-called unemployment volatility puzzle. The important role of mismatch shocks for the natural rate dynamics is a key result of this paper that, we believe, is plausible and intuitive. The mismatch shock captures variations in structural factors (like mismatch, changes in the composition of the unemployment pool, search intensity, and demographic factors, among others) and these factors are the drivers of the natural rate in the spirit of Friedman (1968).

Note, however, that according to our estimates, wage bargaining shocks are almost white noise. This finding is in keeping with the interpretation of wage markup shocks as measurement errors that is favored by Justiniano, Primiceri, and Tambalotti (2013).
This analysis of the natural rate of unemployment has important policy implications, at least if the Fed’s mission is consistent with the view proposed above by Kocherlakota (2011). According to our model, expansionary policies during the Great Recession were justified by an unemployment gap (defined as the difference between the actual rate and the natural rate) that increased from minus 1 percent to 3 percent as we see in Figure 8. All in all, our results are consistent with the view that the large increase in unemployment during the Great Recession is largely due to cyclical factors whereas structural factors have contributed only to some extent. Nevertheless, negative matching efficiency shocks play a larger role in recent years in slowing down the recovery (see Figures 4 and 5) and in closing the unemployment gap which is almost at zero at the end of our sample in 2013:Q2 (see Figure 8).

6 Sensitivity Analysis

We now evaluate the robustness of our results by considering some extensions that we summarize in Figure 9 where we plot a counterfactual historical decomposition for unemployment over the period 2008:Q1-2013Q2 in the absence of mismatch shocks. We compare these extensions to our baseline model (thin solid line) and to the data (bold solid line).

Different sample period. In the first set of experiments we change the sample period used for estimation. We first extend it until 2013:Q2, to use information on the recent Beveridge curve’s shift for estimation purposes, and then we limit it to the Great Moderation period (1984:Q1-2008:Q3) to rule out structural breaks’ concerns. In both cases all our results are confirmed. Mismatch shocks are slightly more important when the sample period for estimation is limited to the Great Moderation period (purple solid line).

Alternative calibration. In the second set of experiments we change the calibrated value for the elasticity of the matching function to unemployment ($\sigma$) and we reestimate the model over the longest sample (1957Q1-2013Q2). We consider two values at the extremes in the range of recent estimates ($\sigma = 0.55$ and $\sigma = 0.75$). While all our results are broadly confirmed, the calibration of $\sigma$ matters for the importance of mismatch shocks in generating the shift in the Beveridge curve. With $\sigma$ equal to 0.55, the matching efficiency process becomes more counter-cyclical at business cycle frequencies and mismatch shocks are less important in explaining the.

6Several additional figures related to the sensitivity analysis and a more detailed discussion are reported in the online appendix.
recent shift in the Beveridge curve and unemployment dynamics in recent years (dotted line in Figure 9). With $\sigma$ equal to 0.75, matching efficiency always declines in recessions and mismatch shocks are more important to explain unemployment dynamics (red dashed line in Figure 9).

**Time-varying separation rate.** In the third set of experiments we consider shocks to the separation rate. In principle an exogenous increase in the separation rate could also shift the Beveridge curve (Shimer 2005, among others). We reestimate our model by including separation shocks rather than mismatch shocks. The estimate of the separation shock series is almost the mirror image of the matching efficiency series, reflecting the fact that the two shocks are nearly observationally equivalent. The estimate for the natural rate of unemployment does not change significantly and separation shocks are now the dominant drivers of the natural rate. The same shift in the Beveridge curve can be explained either ways, through a decline in matching efficiency or through an increase in the separation rate (cf. Figures in the online Appendix). However, this last explanation is in contrast with the JOLTS data. In fact, somewhat surprisingly, the separation rate is constant throughout the Great Recession and declines afterwards. Layoffs increased sharply during the crisis but quits declined by a comparable amount leaving the aggregate separation rate almost unaffected. In more recent years layoffs declined to pre-crisis levels while the increase in quits has been moderate, thus leaving the separation rate still well below its pre-crisis value (cf. Christiano, Eichenbaum and Trabandt 2014; Lubik, 2013). The discrepancy between the JOLTS data and the estimated separation shock suggests that, at least during the Great Recession, other factors than separation were at work and contributed to the shift in the Beveridge curve.

We provide two last experiments to try to overcome the near observational equivalence between mismatch and separation shocks. In the first case we relax the assumption of constant job separation rate without introducing any additional exogenous shock. The separation rate is now a function of the exogenous technology and investment-specific shocks, the two main sources of business cycle fluctuations in our model and we assume that the separation rate is negatively correlated with the state of the economy. Mismatch shocks become slightly less important for unemployment dynamics (grey squared line in Figure 9) but broadly speaking all our main results are confirmed also in this case in which the separation rate reacts to the state of the economy. In our last experiment we use data on the separation rate from JOLTS for the period 2000:Q4-2013:Q2 as an observable variable in the estimation. The separation rate still reacts to the state of the economy but we include also an exogenous component (i.e. a shock to
the separation rate). Therefore, mismatch shocks coexist with separation shocks, unlike in the previous case. In this set-up mismatch shocks are still important for unemployment dynamics in recent years but slightly less than in our baseline (blue dotted line in Figure 9). We find once again evidence of a material deterioration in matching efficiency during and after the Great Recession. Moreover, the mismatch shock is still the main driver of the natural rate, although the role of the separation shock is non negligible. Finally, mismatch shocks are still important to explain the shift to the right of the Beveridge curve since separation shocks tend to shift the curve to the left over the period 2009-2012.

7 Conclusion

In this paper we identify a substantial decline in matching efficiency during the Great Recession and we investigate the macroeconomic consequences of this phenomenon in the context of a New Keynesian model with search and matching frictions extended with matching efficiency shocks and a generalized hiring cost function. We find that the estimated decline in matching efficiency raises the actual unemployment rate by around 1.3 percentage points and the natural rate by 2 percentage points during the Great Recession. In normal times mismatch shocks are almost irrelevant for business cycle fluctuations but, nevertheless, these can play a role in periods when unemployment and vacancies comove. Our key result is that mismatch shocks are the dominant driver of the natural rate and are thus crucial to obtain a reliable estimate of it. Ignoring mismatch shocks, as in a large part of the previous literature, is perhaps not crucial along some dimensions (Christiano, Eichenbaum and Trabandt, 2014) but is definitely problematic when considering the natural rate and the related policy implications.

Finally, we want to underscore that matching efficiency shocks have a broad interpretation. We see them as catch-all disturbances that soak up changes in various features of the aggregate labor market, not only mismatch. Like the Solow residual of the neo-classical production function, matching efficiency is likely to incorporate a non negligible endogenous component. For example, search intensity by workers and firms may play a nontrivial role, as does variable capacity utilization in the production function. Our paper is only a first step in the identification of structural factors in the labor market. More generally, we believe there is scope for future research on how to “purify” the matching function’s Solow residual, as has been done for the production function (cf. for recent advances Barnichon and Figura 2014; Borowczyk-Martins, Jolivet, and Postel-Vinay 2013; and Sedlacek 2014).
References


Table 1: Calibrated Parameters

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
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<td>Capital depreciation rate</td>
<td>$\delta$ 0.0250</td>
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<tr>
<td>Capital share</td>
<td>$\alpha$ 0.33</td>
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<td>Elasticity of substitution btw goods</td>
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<td>Backward-looking price setting</td>
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<td>Replacement rate</td>
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<td>Job destruction rate</td>
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<tr>
<td>Elasticity of matches to unemp.</td>
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<td>Probability to fill a vacancy within a quarter</td>
<td>$q$ 0.70</td>
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<tr>
<td>Exogenous spending/output ratio</td>
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<tr>
<td>Unemployment rate</td>
<td>$U$ 0.0578</td>
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<tr>
<td>Quarterly gross growth rate</td>
<td>$z$ 1.0039</td>
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<tr>
<td>Quarterly gross inflation rate</td>
<td>$\pi$ 1.0088</td>
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<td>Quarterly gross nominal interest rate</td>
<td>$R$ 1.0139</td>
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Table 2: Priors and Posteriors of Structural Parameters

<table>
<thead>
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<th>Parameter</th>
<th>Priors</th>
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<th>Median</th>
<th>95%</th>
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<tr>
<td>Weight of pre-match cost in total hiring cost</td>
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<td>0.35</td>
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<td>Habit in consump.</td>
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<td>0.65</td>
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<td>Invest. adj. cost</td>
<td>$\phi_I$</td>
<td>IGamma (5,1)</td>
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<td>Capital ut. cost</td>
<td>$\phi_u$</td>
<td>IGamma (0.5,0.1)</td>
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<td>Price adjust. cost</td>
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<td>Wage adjust. cost</td>
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<td>IGamma (150,25)</td>
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<td>Wage indexation</td>
<td>$\rho$</td>
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<td>Interest smoothing</td>
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<td>Resp. to inflation</td>
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<td>Resp. to growth</td>
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Table 3: Priors and Posteriors of Shock Parameters

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<td>Technology growth $\rho_z$</td>
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<td>$100\sigma_z$</td>
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<td>1.18</td>
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<td>Monetary policy $\rho_{mp}$</td>
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<td>$100\sigma_{mp}$</td>
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<td>0.19</td>
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<td>Investment $\rho_\zeta$</td>
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<td>0.96</td>
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<td>$100\sigma_{\sigma^*}$</td>
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Table 4: Variance Decomposition (in %)

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<th>Inflation</th>
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<td>Fiscal</td>
<td>13.9</td>
<td>3.8</td>
<td>9.9</td>
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</table>
Figure 1: Impulse responses to a one-standard deviation negative matching efficiency shock, computed at the posterior mode, and under alternative calibrations of the pre-match hiring cost weight.
Figure 2: Impulse responses to a one-standard-deviation negative matching efficiency shock in the actual economy with nominal rigidities, computed at the posterior mode, and in the counter-factual economy with no nominal rigidities.
Figure 9: Actual Unemp. vs. Counterfactual (No Match Shocks)

- Actual
- Baseline
- 1957-2013
- 1985-2008
- $\sigma = 0.55$
- $\sigma = 0.75$
- TV sep (8 obs)
- TV sep (9 obs)

ppt

08 09 10 11 12 13

0 1 2 3 4 5 6 7 8 9 10
Appendix for
“Mismatch shocks and unemployment during the Great Recession”

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1 Roadmap

Section 2 describes the model. Section 3 deals with our empirical strategy. Section 4 provides additional results from the baseline estimation. Section 5 offers some robustness checks.

2 Model

2.1 The representative family

There is a continuum of identical households of mass one. Each household is a large family, made of a continuum of individuals of measure one. Family members are either working or searching for a job.\(^1\)

Following Merz (1995), we assume that family members pool their income before the head of the family chooses optimally per capita consumption.\(^2\)

The representative family enters each period \(t = 0, 1, 2, \ldots\), with \(B_{t-1}\) bonds and \(K_{t-1}\) units of physical capital. At the beginning of each period, bonds mature, providing \(B_t\) units of money. The representative family uses some of this money to purchase \(B_t\) new bonds at nominal cost \(B_t/R_t\), where \(R_t\) denotes the gross nominal interest rate between period \(t\) and \(t+1\).

The representative household owns capital and chooses the capital utilization rate, \(u_t\), which transforms physical capital into effective capital according to

\[
K_t = u_t K_{t-1}. \tag{1}
\]

The household rents \(K_t(i)\) units of effective capital to intermediate-goods-producing firm \(i \in [0, 1]\) at the nominal rate \(r^K_t\). The household’s choice of \(K_t(i)\) must satisfy

\[
K_t = \int_0^1 K_t(i) \, di. \tag{2}
\]

The cost of capital utilization is \(a(u_t)\) per unit of physical capital. We assume the following functional form for the function \(a\),

\[
a(u_t) = \phi_{u_1} (u_t - 1) + \frac{\phi_{u_2}}{2} (u_t - 1)^2, \tag{3}
\]

and that \(u_t = 1\) in steady state.

Each period, \(N_t(i)\) family members are employed at intermediate goods-producing firm \(i \in [0, 1]\). Each worker employed at firm \(i\) works a fixed amount of hours and earns the nominal wage \(W_t(i)\). \(N_t\) denotes aggregate employment in period \(t\) and is given by

\[
N_t = \int_0^1 N_t(i) \, di. \tag{4}
\]

The remaining \((1 - N_t)\) family members are unemployed and and each receives nominal unemployment benefits \(b_t\), financed through lump-sum taxes.

During period \(t\), the representative household receives total nominal factor payments \(r^K_t K_t + W_t N_t +\)

\(^1\)The model abstracts from the labor force participation decision.

In addition, the household also receives nominal profits $D_t(i)$ from each firm $i \in [0, 1]$, for a total of
\[
D_t = \int_0^1 D_t(i) \, di.
\]
In each period $t = 0, 1, 2, \ldots$ the family uses these resources to purchase finished goods, for both consumption and investment purposes, from the representative finished goods-producing firm at the nominal price $P_t$. The law of motion of physical capital is
\[
K_t \leq (1 - \delta) K_{t-1} + \mu_t \left[ 1 - \mathcal{L} \left( \frac{I_t}{I_{t-1}} \right) \right] I_t,
\]
where $\delta$ denotes the depreciation rate. The function $\mathcal{L}$ captures the presence of adjustment costs in investment, as in Christiano, Eichenbaum and Evans (2005). We assume the following functional form for the function $\mathcal{L}$,
\[
\mathcal{L} \left( \frac{I_t}{I_{t-1}} \right) = \frac{\phi_I}{2} \left( \frac{I_t}{I_{t-1}} - gt \right)^2,
\]
where $g_I$ is the steady-state growth rate of investment. Hence, along the balanced growth path, $\mathcal{L}(g_I) = \mathcal{L}'(g_I) = 0$ and $\mathcal{L}''(g_I) = \phi_I > 0$. $\mu_t$ is an investment-specific technology shock affecting the efficiency with which consumption goods are transformed into capital. The investment-specific shock follows the exogenous stationary autoregressive process
\[
\ln \mu_t = \rho_{\mu} \ln \mu_{t-1} + \varepsilon_{\mu t},
\]
where $\varepsilon_{\mu t}$ is i.i.d. $N(0, \sigma_{\mu}^2)$.

The family's budget constraint is given by
\[
P_tC_t + P_tI_t + \frac{B_t}{\varepsilon_{bt}R_t} \leq B_{t-1} + W_t N_t + (1 - N_t) b_t + r^K_t u_t \bar{K}_{t-1}
\]
\[
- P_t a(u_t) \bar{K}_{t-1} - T_t + D_t
\]
for all $t = 0, 1, 2, \ldots$ As in Smets and Wouters (2007), the shock $\varepsilon_{bt}$ drives a wedge between the central bank's instrument rate $R_t$ and the return on assets held by the representative family. As noted by De Graeve, Emiris and Wouters (2009), this disturbance works as an aggregate demand shock and generates a positive comovement between consumption and investment. The risk-premium shock $\varepsilon_{bt}$ follows the autoregressive process
\[
\ln \varepsilon_{bt} = \rho_b \ln \varepsilon_{bt-1} + \varepsilon_{bt},
\]
where $0 < \rho_b < 1$, and $\varepsilon_{bt}$ is i.i.d. $N(0, \sigma_b^2)$.

The family's lifetime utility is described by
\[
E_t \sum_{s=0}^{\infty} \beta^s \ln (C_{t+s} - hC_{t+s-1})
\]
where $0 < \beta < 1$. When $h > 0$, the model allows for habit formation in consumption and consumption responds gradually to shocks.

The head of the family chooses $C_t$, $B_t$, $u_t$, $I_t$, and $\bar{K}_t$ for each $t = 0, 1, 2, \ldots$ to maximize the expected
lifetime utility (10) subject to the constraints (6) and (9).

The Lagrangean reads

$$E_0 \sum_{t=0}^{\infty} \left\{ \beta^t \ln \left( C_t - hC_{t-1} \right) + \beta^t \lambda_t \left[ \frac{B_{t-1} + W_t N_t + (1 - N_t)b_t + r^K u_t K_{t-1} - T_t + D_t}{P_t} - a(u_t) K_{t-1} - C_t - I_t - \frac{B_t}{\epsilon_B R_t P_t} \right] \right\}$$

$$+ \beta^t Y_t \left[ (1 - \delta) K_{t-1} + \mu_t \left( 1 - \frac{\phi_1}{2} \left( \frac{I_t}{I_{t-1}} - g_I \right)^2 \right) I_t - K_t \right]$$

(12)

The first order conditions for this problem are

- $C_t$:
  $$\lambda_t = \frac{1}{C_t - hC_{t-1}} - \beta h E_t \left( \frac{1}{C_{t+1} - hC_t} \right)$$
  (13)

- $B_t$:
  $$\lambda_t = \epsilon_B R_t \beta E_t \left( \frac{\lambda_{t+1}}{P_{t+1}} \right)$$
  (14)

- $u_t$:
  $$(\phi_{u1} - \phi_{u2}) + \phi_{u2} u_t = \tilde{r}_t^K$$
  (15)

  where $\tilde{r}_t^K$ denotes the real rental rate of capital $\tilde{r}_t^K = r_t^K / P_t$.

- $I_t$:
  $$1 = u_t \mu_t \left[ 1 - \frac{\phi_1}{2} \left( \frac{I_t}{I_{t-1}} - g_I \right)^2 - \phi_1 \left( \frac{I_t}{I_{t-1}} - g_I \right) \left( \frac{I_t}{I_{t-1}} \right) \right] + \beta E_t u_{t+1} \mu_{t+1} - \lambda_{t+1} - \lambda_t \phi_1 \left( \frac{I_{t+1}}{I_t} - g_I \right) \left( \frac{I_{t+1}}{I_t} \right)^2$$
  (16)

  where $u_t$ is the marginal Tobin’s Q: the Lagrange multiplier associated with the investment adjustment constraint, $Y_t$, normalized by $\lambda_t$.

- $K_t$:
  $$u_t = \beta E_t \left\{ \frac{\lambda_{t+1}}{\lambda_t} \left[ (1 - \delta) u_{t+1} + \tilde{r}_t^K u_{t+1} - a(u_{t+1}) \right] \right\}$$
  (17)

- $\Lambda_t$:
  $$\frac{B_{t-1} + W_t N_t + (1 - N_t) b_t + r^K u_t K_{t-1} - T_t + D_t}{P_t} - a(u_t) K_{t-1} = C_t + I_t + \frac{B_t}{\epsilon_B R_t P_t}$$
  (18)

  where $\Lambda_t$ denotes the multiplier on (9) and can be interpreted as the utility to the household of an additional unit of wealth at date $t$.

- $Y_t$:
  $$K_t = (1 - \delta) K_{t-1} + \mu_t \left[ 1 - \frac{\phi_1}{2} \left( \frac{I_t}{I_{t-1}} - g_I \right)^2 \right] I_t$$
  (19)
where \( T_t \) denotes the multiplier on (6) and can be interpreted as the utility to the household of an additional unit of physical capital at date \( t \).

### 2.2 The representative finished goods-producing firm

During each period \( t = 0, 1, 2, \ldots \), the representative finished goods-producing firm uses \( Y_t(i) \) units of each intermediate good \( i \in [0, 1] \), purchased at the nominal price \( P_t(i) \), to manufacture \( Y_t \) units of the finished good according to the constant-returns-to-scale technology described by

\[
\left[ \int_0^1 Y_t(i)^{(\theta_t-1)/\theta_t} \, di \right]^{\theta_t/(\theta_t-1)} \geq Y_t, \tag{20}
\]

where \( \theta_t \) translates into a random shock to the markup of price over marginal cost. This markup shock follows the autoregressive process

\[
\ln \theta_t = (1 - \rho_\theta) \ln \theta + \rho_\theta \ln \theta_{t-1} + \varepsilon_\theta_t, \tag{21}
\]

where \( 0 < \rho_\theta < 1, \theta > 1, \) and \( \varepsilon_\theta_t \) is i.i.d. \( N(0, \sigma_\theta^2) \).

Intermediate good \( i \) sells at the nominal price \( P_t(i) \), while the finished good sells at the nominal price \( P_t \). Given these prices, the finished goods-producing firm chooses \( Y_t \) and \( Y_t(i) \) for all \( i \in [0, 1] \) to maximize its profits

\[
P_t Y_t - \int_0^1 P_t(i) Y_t(i) \, di, \tag{22}
\]

subject to the constraint (17) for each \( t = 0, 1, 2, \ldots \). The first-order conditions for this problem are (17) with equality and

\[
Y_t(i) = \left[ \frac{P_t(i)}{P_t} \right]^{-\theta_t} Y_t \tag{23}
\]

for all \( i \in [0, 1] \) and \( t = 0, 1, 2, \ldots \).

Competition in the market for the finished good drives the finished goods-producing firm’s profits to zero in equilibrium. This zero-profit condition determines \( P_t \) as

\[
P_t = \left[ \int_0^1 P_t(i)^{1-\theta_t} \, di \right]^{1/(1-\theta_t)} \tag{24}
\]

for all \( t = 0, 1, 2, \ldots \).

### 2.3 The representative intermediate goods-producing firm

Each intermediate goods-producing firm \( i \in [0, 1] \) enters in period \( t \) with a stock of \( N_{t-1}(i) \) employees carried from the previous period. At the beginning of period \( t \), before production starts, \( \rho N_{t-1}(i) \) jobs are destroyed, where \( \rho \) is the exogenous job destruction rate. The pool of workers \( \rho N_{t-1} \) who have lost their job at the beginning of period \( t \) start searching immediately and can possibly be hired in period \( t \). The number of employees at firm \( i \) evolves according to

\[
N_t(i) = (1 - \rho) N_{t-1}(i) + m_t(i). \tag{25}
\]
\( m_t(i) \) denotes the flow of new employees hired by firm \( i \) in period \( t \), and is given by
\[
 m_t(i) = q_t V_t(i),
\]
(26)
where \( V_t(i) \) denotes vacancies posted by firm \( i \) in period \( t \) and \( q_t \) is the aggregate probability of filling a vacancy in period \( t \). Workers hired in period \( t \) take part to period \( t \) production. Employment is therefore an \emph{instantaneous} margin. However, each period some vacancies and job seekers remain unmatched. As a consequence, a firm-worker pair enjoys a joint surplus that motivates the existence of a long-run relationship between the two parties.

Aggregate employment \( N_t = \int_0^1 N_t(i) \, di \) evolves over time according to
\[
 N_t = (1 - \rho) N_{t-1} + m_t,
\]
(27)
where \( m_t = \int_0^1 m_t(i) \, di \) denotes aggregate matches in period \( t \). Similarly, the aggregate vacancies is equal to \( V_t = \int_0^1 V_t(i) \, di \). The pool of job seekers in period \( t \), denoted by \( S_t \), is given by
\[
 S_t = 1 - (1 - \rho) N_{t-1}.
\]
(28)

The matching process is described by the following aggregate CRS function
\[
 m_t = \zeta_t S_t^\sigma V_t^{1-\sigma},
\]
(29)
where \( \zeta_t \) is an exogenous disturbance to the efficiency of the matching technology. We label this disturbance the mismatch shock and assume it follows the exogenous stationary stochastic process
\[
 \ln \zeta_t = (1 - \rho_\zeta) \ln \zeta + \rho_\zeta \ln \zeta_{t-1} + \varepsilon_{\zeta t},
\]
(30)
where \( \zeta > 0 \) denotes the steady-state efficiency of the matching technology and \( \varepsilon_{\zeta t} \) is \emph{i.i.d.} \( N \left( 0, \sigma_\zeta^2 \right) \). The probability \( q_t \) to fill a vacancy in period \( t \) is given by
\[
 q_t = \frac{m_t}{V_t} = \zeta_t \Theta_t^{-\sigma},
\]
(31)
where \( \Theta_t \) denotes the tightness of the labor market \( \Theta_t = V_t/S_t \). The probability \( s_t \) for a job seeker to find a job is
\[
 s_t = \frac{m_t}{S_t} = \zeta_t \Theta_t^{1-\sigma}.
\]
(32)

Finally aggregate unemployment is defined by \( U_t = 1 - N_t \).

During each period \( t = 0, 1, 2, \ldots \), the representative intermediate goods-producing firm combines \( N_t(i) \) homogeneous employees with \( K_t(i) \) units of efficient capital to produce \( Y_t(i) \) units of intermediate good \( i \) according to the constant-returns-to-scale technology described by
\[
 Y_t(i) = A_t^{1-\alpha} K_t(i)^\alpha N_t(i)^{1-\alpha}.
\]
(33)
\( A_t \) is an aggregate labor-augmenting technology shock whose growth rate, \( z_t = A_t/A_{t-1} \), follows the exogenous stationary stochastic process
\[
 \ln z_t = (1 - \rho_z) \ln z + \rho_z \ln z_{t-1} + \varepsilon z_t,
\]
(34)

\[ 6 \]
where \( z > 1 \) denotes the steady-state growth rate of the economy and \( \varepsilon_{zt} \) is i.i.d. \( N(0, \sigma^2_z) \).

The firm faces costs of hiring workers. As in Yashiv (2000 and 2006), hiring costs are a convex function of the linear combination of the number of vacancies and the number of hires. Hiring costs are measured in terms of aggregate output, and given by

\[
\frac{\kappa}{2} \left( \phi_V V_t(i) + (1 - \phi_V) q_t V_t(i) \right)^2 Y_t, \tag{35}
\]

where \( \phi_V \) governs the magnitude of these costs.\(^3\)

Intermediate goods substitute imperfectly for one another in the production function of the representative finished goods-producing firm. Hence, each intermediate goods-producing firm \( i \in [0, 1] \) sells its output \( Y_t(i) \) in a monopolistically competitive market, setting \( P_t(i) \), the price of its own product, with the commitment of satisfying the demand for good \( i \) at that price. Firms take the nominal wage as given when maximizing the discounted value of expected future profits.

Each intermediate goods-producing firm faces costs of adjusting its nominal price between periods (Rotemberg 1982), measured in terms of the finished good and given by

\[
\frac{\phi_P}{2} \left( \frac{P_t(i)}{\pi^{1-\varsigma}_{t-1} P_{t-1}(i)} - 1 \right)^2 Y_t, \tag{36}
\]

\( \phi_P \) governs the magnitude of the price adjustment cost. \( \pi_t = \frac{P_t}{P_{t-1}} \) denotes the gross rate of inflation in period \( t \). \( \pi > 1 \) denotes the steady-state gross rate of inflation and coincides with the central bank’s target. The parameter \( 0 \leq \varsigma \leq 1 \) governs the importance of backward-looking behavior in price setting (cf. Ireland 2007).

Following Arsenau and Chugh (2008), firms face quadratic wage-adjustment costs which are proportional to the size of their workforce and measured in terms of the finished good

\[
\frac{\phi_W}{2} \left( \frac{W_t(i)}{z \pi^{\varrho}_{t-1} \pi^{1-\varsigma}_{t-1} W_{t-1}(i)} - 1 \right)^2 N_t(i) Y_t, \tag{37}
\]

where \( \phi_W \geq 0 \) governs the magnitude of the wage adjustment cost. The parameter \( 0 \leq \varrho \leq 1 \) governs the importance of backward-looking behavior in wage setting.

Adjustment costs on the hiring rate, price and wage changes make the intermediate goods-producing firm’s problem dynamic. It chooses \( K_t(i), N_t(i), V_t(i) \) and \( Y_t(i) \) and \( P_t(i) \) for all \( t = 0, 1, 2, \ldots \) to maximize its total market value, given by

\[
E_t \sum_{s=0}^{\infty} \beta^s \Lambda_{t+s} \left( \frac{D_{t+s}(i)}{P_{t+s}} \right), \tag{38}
\]

where \( \beta^s \Lambda_t/P_t \) measures the marginal utility to the representative household of an additional dollar of profits during period \( t \) and where

\[
D_t(i) = P_t(i) Y_t(i) - W_t(i) N_t(i) - r_t^K K_t(i) - \frac{\kappa}{2} \left( \phi_V V_t(i) + (1 - \phi_V) q_t V_t(i) \right)^2 P_t Y_t
- \frac{\phi_P}{2} \left( \frac{P_t(i)}{\pi^{1-\varsigma}_{t-1} P_{t-1}(i)} - 1 \right)^2 P_t Y_t
- \frac{\phi_W}{2} \left( \frac{W_t(i)}{z \pi^{\varrho}_{t-1} \pi^{1-\varsigma}_{t-1} W_{t-1}(i)} - 1 \right)^2 N_t(i) P_t Y_t, \tag{39}
\]

\(^3\)Hiring costs are proportional to output and thus inherit the common stochastic trend driving productivity. This specification ensures that the unemployment rate remains stationary along the balanced steady-state growth path.
subject to the constraints

\[ Y_t(i) = \left[ \frac{P_t(i)}{P_t} \right]^{-\theta_t} Y_t, \quad (40) \]
\[ Y_t(i) \leq K_t(i)^\alpha [A_tN_t(i)]^{1-\alpha}, \quad (41) \]
\[ N_t(i) = \chi N_{t-1}(i) + q_t V_t(i), \quad (42) \]

where \( \chi \equiv 1 - \rho \) is the job survival rate.

This problem is equivalent to the one of choosing \( K_t(i), N_t(i), V_t(i) \) and \( P_t(i) \) to maximize (35), where

\[ \frac{D_t(i)}{P_t} = \left( \frac{P_t(i)}{P_t} \right)^{1-\theta_t} Y_t \left( \frac{W_t(i) N_t(i) + r_t^K K_t(i)}{P_t} \right) - \frac{\kappa}{2} \left( \frac{\phi_V V_t(i) + (1 - \phi_V) q_t V_t(i)}{N_t(i)} \right)^2 Y_t \]
\[ - \frac{\phi_P}{2} \left( \frac{P_t(i)}{\pi_{t-1}^\alpha P_{t-1}(i) - 1} \right)^2 Y_t - \frac{\phi_W}{2} \left( \frac{W_t(i)}{z \pi_{t-1}^\alpha W_{t-1}(i) - 1} \right)^2 N_t(i) Y_t, \quad (43) \]

subject to the constraints

\[ \left[ \frac{P_t(i)}{P_t} \right]^{-\theta_t} Y_t \leq K_t(i)^\alpha [A_tN_t(i)]^{1-\alpha}, \quad (44) \]
\[ N_t(i) = \chi N_{t-1}(i) + q_t V_t(i), \quad (45) \]

for all \( t = 0, 1, 2, \ldots \).

The Lagrangean reads

\[ E_0 \sum_{t=0}^\infty \beta^t \beta^t N_t \left[ \left( \frac{P_t(i)}{P_t} \right)^{1-\theta_t} Y_t \left( \frac{W_t(i) N_t(i) + r_t^K K_t(i)}{P_t} \right) - \frac{\kappa}{2} \left( \frac{\phi_V V_t(i) + (1 - \phi_V) q_t V_t(i)}{N_t(i)} \right)^2 Y_t \right] \]
\[ - \frac{\phi_P}{2} \left( \frac{P_t(i)}{\pi_{t-1}^\alpha P_{t-1}(i) - 1} \right)^2 Y_t - \frac{\phi_W}{2} \left( \frac{W_t(i)}{z \pi_{t-1}^\alpha W_{t-1}(i) - 1} \right)^2 N_t(i) Y_t \]
\[ + E_0 \sum_{t=0}^\infty \beta^t \Psi_t(i) \left[ \chi N_{t-1}(i) + q_t V_t(i) - N_t(i) \right] + E_0 \sum_{t=0}^\infty \beta^t \Xi_t(i) \left[ K_t(i)^\alpha (A_tN_t(i))^{1-\alpha} - \left( \frac{P_t(i)}{P_t} \right)^{-\theta_t} Y_t \right] \]

The multiplier \( \Psi_t(i) \) measures the value to firm \( i \), expressed in utils, of an additional job in period \( t \). The multiplier \( \Xi_t(i) \) measures the value to firm \( i \), expressed in utils, of an additional unit of output in period \( t \). Hence, \( \xi_t(i) = \Xi_t(i) / \Lambda_t \) represents firm \( i \)’s real marginal cost in period \( t \).

The first-order conditions for this problem are

- \( K_t(i) \):
  \[ \hat{\gamma}_t^K = \xi_t(i) \alpha K_t(i)^{\alpha-1} (A_tN_t(i))^{1-\alpha} \quad (47) \]

- \( N_t(i) \):
  \[ \frac{\Psi_t(i)}{\Lambda_t} = \xi_t(i) (1 - \alpha) \frac{Y_t(i)}{N_t(i)} - \frac{W_t(i)}{P_t} - \frac{\phi_W}{2} \left( \frac{W_t(i)}{z \pi_{t-1}^\alpha W_{t-1}(i) - 1} \right)^2 Y_t \]
  \[ + \frac{\kappa}{N_t(i)} \left[ \frac{\phi_V V_t(i) + (1 - \phi_V) q_t V_t(i)}{N_t(i)} \right]^2 Y_t + \beta \chi \frac{\Lambda_{t+1} \Psi_{t+1}(i)}{\Lambda_t \Lambda_{t+1}} \quad (48) \]

This condition tells that the costs and benefits of hiring an additional worker must be equal.

- \( V_t(i) \):
  \[ \frac{\Psi_t(i)}{\Lambda_t} = \left( \frac{\phi_V + (1 - \phi_V) q_t}{N_t(i)} \right)^2 \frac{\kappa \gamma_t N_t(i)}{q_t} \quad (49) \]
• Vacancy posting condition:

\[
\left(\frac{\phi_V + (1 - \phi_V) q_t}{N_t(i)}\right)^2 \frac{\kappa Y_t(i)}{q_t} = \xi_t(i) (1 - \alpha) \frac{Y_t(i)}{N_t(i)} - \frac{W_t(i)}{P_t} - \frac{\phi_W}{2} \left(\frac{W_t(i)}{z_{t-1} \pi^{1-\theta W_{t-1}(i)} - 1}\right)^2 Y_t
\]

\[+ \frac{\kappa}{N_t(i)} \left[\frac{\phi_V V_t(i) + (1 - \phi_V) q_t V_t(i)}{N_t(i)}\right]^2 Y_t
\]

\[+ \beta \chi A_{t+1} I_t \left(\phi_V + (1 - \phi_V) q_{t+1}\right) \frac{2 \kappa Y_{t+1} V_{t+1}(i)}{q_{t+1}} \]

\[ (50) \]

• \( P_t(i) \):

\[(1 - \theta_t) \left(\frac{P_t(i)}{P_t}\right)^{-\theta_t} = \phi_p \left(\frac{P_t(i)}{\pi^{1-\theta P_{t-1}(i)} - 1}\right) \left(\frac{P_t}{\pi^{1-\theta P_{t-1}(i)} - 1}\right) - \theta_t \xi_t(i) \left(\frac{P_t(i)}{P_t}\right)^{(1+\theta_t)}
\]

\[ - \beta \phi_p E_t \left[\frac{A_{t+1} I_t}{\pi^{1-\theta P_{t+1}(i)} - 1}\right] \left(\frac{P_{t+1}(i)}{\pi^{1-\theta P_{t+1}(i)} - 1}\right) Y_{t+1} \left(\frac{P_{t+1}(i)}{P_t(i)}\right) \]

\[ (51) \]

• \( \Psi_t(i) \):

\[ N_t(i) = \chi N_{t-1}(i) + q_t V_t(i) \]

\[ (52) \]

• \( \Xi_t(i) \):

\[ A_t^{1-\alpha} K_t(i)^{\alpha} N_t(i)^{1-\alpha} = \left(\frac{P_t(i)}{P_t}\right)^{-\theta_t} Y_t \]

\[ (53) \]

### 2.4 Wage setting

Each period, intermediate-good producing firm \( i \) bargains with each of its employees individually over the nominal wage \( W_t(i) \) to maximize the match surplus according to Nash bargaining,

\[ W_t(i) = \arg \max \left[\Delta_t(i)^{\eta_t} J_t(i)^{1-\eta_t}\right]. \]

\[ (54) \]

\( \Delta_t(i) \) denotes the surplus of the representative worker while \( J_t(i) \) denotes the surplus of the firm. Both \( \Delta_t(i) \) and \( J_t(i) \) are expressed in real terms. \( \eta_t \) denotes the worker’s bargaining power which evolves exogenously according to

\[ \ln \eta_t = (1 - \rho_{\eta}) \ln \eta + \rho_{\eta} \ln \eta_{t-1} + \varepsilon_{\eta t}, \]

\[ (55) \]

where \( 0 < \eta < 1 \) and \( \varepsilon_{\eta t} \) is i.i.d. \( N(0, \sigma_{\eta}^2) \).

The family’s value function is given by

\[ \Omega(N_t) = \ln (C_t - h C_{t-1}) + \lambda_t \left[\int_0^1 \frac{1}{P_t} \frac{W_t(i) N_t(i)}{P_t} di + (1 - N_t(i)) \left(\frac{\eta_t}{P_t}\right) \right] + \lambda_t \left[D_t - a(u_t) K_{t-1} - \frac{B_t}{\varepsilon_t R_t P_t}\right]
\]

\[ + \gamma_t \left[(1 - \delta) K_{t-1} + \mu_t \left(1 - \phi_t \left(\frac{I_t}{I_{t-1}} - g_t\right)^2\right) I_t - K_t\right] + \beta E_t \Omega(N_{t+1}). \]

\[ (56) \]
Following Trigari (2009) and Ravenna and Walsh (2008), the value of the nominal wage along the balanced growth path to ensure that the model is consistent with balanced growth, unemployment benefits are proportional to the value of the nominal wage along the balanced growth path \( b_t = \tau W_{as,t} \), where \( \tau \) is the replacement ratio. Following Trigari (2009) and Ravenna and Walsh (2008), \( \Delta_t (i) \) is defined as the change in the family’s value function \( \Omega (N_t) \) from having one additional member employed. Thus, the surplus of an employee at firm \( i \), expressed in utils, is given by

\[
\Delta_t (i) = \Delta_t \left( \frac{W_t (i)}{P_t} - \frac{b_t}{P_t} \right) + \beta E_t [\chi (1 - s_{t+1})] \Delta_{t+1} (i).
\]

The worker’s surplus from the match, expressed in consumption goods, is given by

\[
\Delta_t (i) = \frac{W_t (i)}{P_t} - \frac{b_t}{P_t} + \beta E_t [\chi (1 - s_{t+1})] \left( \frac{\Delta_{t+1}}{\Delta_t} \right) \Delta_{t+1} (i).
\]

The employer’s surplus from the match, expressed in real terms, is given by \( J_t (i) = \frac{\Psi_t (i)}{\Lambda_t} \)

\[
J_t (i) = \xi_t (i) (1 - \alpha) \frac{Y_t (i)}{N_t (i)} - \frac{W_t (i)}{P_t} - \frac{\phi_W}{2} \left( \frac{W_t (i)}{z_{\pi_{t-1}^{t}}} - \pi_{t-1}^{t} \right) - 1 \right)^2 \frac{Y_t}{\Lambda_t}.
\]

Nash bargaining over the nominal wage yields the following first-order condition

\[
\eta_t J_t (i) \frac{\partial \Delta_t (i)}{\partial W_t (i)} = - (1 - \eta_t) \Delta_t (i) \frac{\partial J_t (i)}{\partial W_t (i)},
\]

where

\[
\frac{\partial \Delta_t (i)}{\partial W_t (i)} = \frac{1}{P_t},
\]

\[
\frac{\partial J_t (i)}{\partial W_t (i)} = \left\{ \frac{1}{P_t} + \phi_W Y_t \left( \frac{1}{z_{\pi_{t-1}^{t}}} \right) \left( \frac{1}{z_{\pi_{t-1}^{t}}} - \pi_{t-1}^{t} \right) + \beta \phi_W E_t \right\}.
\]

When \( \phi_W = 0 \), adjusting nominal wages is costless for the firm. In that case, the effects of a marginal increase in the nominal wage on the worker’s surplus and on the firm’s surplus have the same magnitude (with opposite signs):

\[
\text{if } \phi_W = 0, \text{ then } \frac{\partial \Delta_t (i)}{\partial W_t (i)} = - \frac{\partial J_t (i)}{\partial W_t (i)} = \frac{1}{P_t}.
\]

In the absence of nominal wage-adjustment costs, Nash bargaining over the nominal wage implies the usual first-order condition

\[
\Delta_t (i) = \left( \frac{\eta_t}{1 - \eta_t} \right) J_t (i).
\]

Thus, as pointed out by Arsenau and Chugh (2008), Nash bargaining over the nominal wage when there are no nominal wage adjustment costs is equivalent to Nash bargaining over the real wage. The presence of nominal wage-adjustment costs (borne by the firm) affects the effective bargaining powers of the firm and the worker respectively. In the presence of nominal wage adjustment costs, the first-order condition from
Nash bargaining is given by
\[
\Delta_t(i) = \frac{\eta_t}{(1-\eta_t)} \left[ \frac{\partial \Delta_t(i)}{\partial W_t(i)} \right] J_t(i),
\]
\[
\Delta_t(i) = \Omega dt J_t(i),
\]
where we have introduced the notation
\[
\Omega dt = \left( \frac{\eta_t}{1-\eta_t} \right) \left( \frac{\partial \Delta_t(i)}{\partial W_t(i)} \right).
\]

Substituting the expressions of the two partial derivatives into the first-order condition, we obtain
\[
\Omega dt \left[ \xi_t(i) (1-\alpha) \frac{Y_t(i)}{N_t(i)} - \frac{W_t(i)}{P_t} - \frac{\phi_W}{2} \left( \frac{W_t(i)}{z\pi_{t-1}^p\pi^1-eW_{t-1}(i)} - 1 \right) \right] Y_t
\]
\[
+ \Omega dt \left[ \kappa \frac{\phi_V V_t(i) + (1-\phi_V) q_t V_t(i)}{N_t(i)} \right] Y_t
\]
\[
+ \Omega dt \beta \chi E_t \left[ \frac{\Lambda_{t+1}}{\Lambda_t} J_{t+1}(i) \right]
\]
\[
= \frac{W_t(i)}{P_t} - \frac{b_t}{P_t} + \beta \chi E_t \left[ (1 - s_{t+1}) \left( \frac{\Lambda_{t+1}}{\Lambda_t} \right) \Omega dt J_{t+1}(i) \right],
\]

Using the fact that \( \Omega dt J_{t+1}(i) = \Omega dt + J_{t+1}(i) \) in the above equation, we obtain
\[
\Omega dt \left[ \xi_t(i) (1-\alpha) \frac{Y_t(i)}{N_t(i)} - \frac{W_t(i)}{P_t} - \frac{\phi_W}{2} \left( \frac{W_t(i)}{z\pi_{t-1}^p\pi^1-eW_{t-1}(i)} - 1 \right) \right] Y_t
\]
\[
+ \Omega dt \left[ \kappa \frac{\phi_V V_t(i) + (1-\phi_V) q_t V_t(i)}{N_t(i)} \right] Y_t
\]
\[
+ \Omega dt \beta \chi E_t \left[ \frac{\Lambda_{t+1}}{\Lambda_t} J_{t+1}(i) \right]
\]
\[
= \frac{W_t(i)}{P_t} - \frac{b_t}{P_t} + \beta \chi E_t \left[ (1 - s_{t+1}) \left( \frac{\Lambda_{t+1}}{\Lambda_t} \right) \Omega dt J_{t+1}(i) \right],
\]

Now, let us recall the definition of the firm’s surplus
\[
J_t(i) = \frac{\Psi_t(i)}{\Lambda_t} = \left( \frac{\phi_V + (1-\phi_V) q_t}{N_t(i)} \right)^2 \frac{\kappa Y_t V_t(i)}{q_t}.
\]

Using this expression of \( J_{t+1}(i) \), the real-wage equation becomes
\[
\frac{W_t(i)}{P_t} - \Omega dt \left[ \xi_t(i) (1-\alpha) \frac{Y_t(i)}{N_t(i)} - \frac{W_t(i)}{P_t} - \frac{\phi_W}{2} \left( \frac{W_t(i)}{z\pi_{t-1}^p\pi^1-eW_{t-1}(i)} - 1 \right) \right] Y_t
\]
\[
- \Omega dt \left[ \kappa \frac{\phi_V V_t(i) + (1-\phi_V) q_t V_t(i)}{N_t(i)} \right] Y_t
\]
\[
= \Omega dt \beta \chi E_t \left[ \frac{\Lambda_{t+1}}{\Lambda_t} \left( \frac{\phi_V + (1-\phi_V) q_{t+1}}{N_{t+1}(i)} \right) \right] Y_{t+1} V_{t+1}(i)
\]
\[
+ \frac{b_t}{P_t} - \beta \chi E_t \left[ (1 - s_{t+1}) \left( \frac{\Lambda_{t+1}}{\Lambda_t} \right) \Omega dt J_{t+1}(i) \right] \left( \frac{\phi_V + (1-\phi_V) q_{t+1}}{N_{t+1}(i)} \right)^2 \frac{\kappa Y_{t+1} V_{t+1}(i)}{q_{t+1}}.
\]
Finally, the equation governing the dynamics of the real wage at firm \(i\) is given by

\[
\frac{W_t(i)}{P_t} = \left( \frac{\Delta t}{1 + \Delta t} \right) \left[ \xi_t(i) (1 - \alpha) \frac{Y_t(i)}{N_t(i)} - \frac{\beta \omega}{2} \left( \frac{W_t(i)}{\pi_t^{\varepsilon_i \theta_t W_t(i)}} - 1 \right)^2 Y_t \right]
+ \frac{\kappa}{N_t(i)} \left( \frac{\phi V_t(i) + (1 - \phi) q_t V_t(i)}{N_t(i)} \right)^2 Y_t
+ \beta \chi E_t \left( \frac{\Delta t+1}{\Delta t} \right) \left( \frac{\phi V_t + (1 - \phi) q_{t+1}}{N_{t+1}(i)} \right)^2 \kappa Y_{t+1} V_{t+1}(i)
+ \frac{1}{1 + \Delta t} \left[ \frac{b_t}{P_t} - \beta \chi E_t \Delta t+1 (1 - s_{t+1}) \left( \frac{\Lambda t+1}{\Lambda t} \right) \left( \frac{\phi V_t + (1 - \phi) q_{t+1}}{N_{t+1}(i)} \right)^2 \kappa Y_{t+1} V_{t+1}(i) \right].
\]

\[2\)

### 2.5 Government

The central bank adjusts the short-term nominal gross interest rate \(R_t\) by following a Taylor-type rule

\[
\ln \left( \frac{R_t}{R} \right) = \rho_r \ln \left( \frac{R_{t-1}}{R} \right) + (1 - \rho_r) \left\{ \rho_y \ln \left( \frac{P_t/P_{t-4}}{\Pi} \right)^{1/4} + \rho_y \ln \left( \frac{Y_t/Y_{t-4}}{G_y} \right)^{1/4} \right\} + \ln \epsilon_{mpt},
\]

where \(\Pi_t = P_t/P_{t-4}\) and \(G_{yt} = Y_t/Y_{t-4}\) and II and \(G_y\) denote the steady state values of \(\Pi_t\) and \(G_{yt}\) respectively. The degree of interest-rate smoothing \(\rho_r\) and the reaction coefficients \(\rho_y, \rho_y\) are positive. The monetary policy shock \(\epsilon_{mpt}\) follows an AR(1) process

\[
\ln \epsilon_{mpt} = \rho_{mp} \ln \epsilon_{mpt-1} + \epsilon_{mpt},
\]

with \(0 \leq \rho_{mp} < 1\) and \(\epsilon_{mpt} \sim i.i.d. N \left(0, \sigma_{mp}^2\right)\).

The government budget constraint is of the form

\[
P_t G_t + (1 - N_t) b_t = \left( \frac{B_t}{R_t} - B_{t-1} \right) + T_t,
\]

where \(T_t\) denotes total nominal lump-sum transfers. Public spending is an exogenous time-varying fraction of GDP

\[
G_t = \left( 1 - \frac{1}{\epsilon_{gt}} \right) Y_t,
\]

where \(\epsilon_{gt}\) evolves according to

\[
\ln \epsilon_{gt} = (1 - \rho_y) \ln \epsilon_g + \rho_y \ln \epsilon_{gt-1} + \epsilon_{gt},
\]

with \(\epsilon_{gt} \sim i.i.d. N \left(0, \sigma_{g}^2\right)\).

### 2.6 The aggregate resource constraint

In a symmetric equilibrium, all intermediate goods-producing firms make identical decisions, so that \(Y_t(i) = Y_t, P_t(i) = P_t, N_t(i) = N_t, V_t(i) = V_t, K_t(i) = K_t\) for all \(i \in [0, 1]\) and \(t = 0, 1, 2, \ldots\). Moreover, workers are homogeneous and all workers at a given firm \(i\) receive the same nominal wage \(W_t(i)\), so that \(W_t(i) = W_t\) for all \(i \in [0, 1]\) and \(t = 0, 1, 2, \ldots\). The aggregate resource constraint is obtained by aggregating the household budget constraint over all intermediate sectors \(i \in [0, 1]\); 

\[
\left[ \frac{1}{\epsilon_{gt}} - \frac{\kappa}{2} \left( \frac{\phi V_t + (1 - \phi) q_t V_t}{\pi_t^{\varepsilon_i \theta_t W_t(i)}} \right)^2 - \frac{\beta \omega}{2} \left( \frac{\pi_t^{\varepsilon_i \theta_t W_t(i)}}{\pi_{t-1}^{\varepsilon_i \theta_t W_{t-1}(i)} - 1} \right)^2 - \frac{\phi V_t}{2} \left( \frac{W_t(i)}{\pi_t^{\varepsilon_i \theta_t W_t(i)}} - 1 \right)^2 \right] Y_t = C_t + I_t + a (u_t) \overline{K}_{t-1}.
\]

(78)
The symmetric equilibrium

In a symmetric equilibrium, \( Y_t(i) = Y_t \), \( P_t(i) = P_t \), \( N_t(i) = N_t \), \( V_t(i) = V_t \), \( K_t(i) = K_t \), \( W_t(i) = W_t \) for all \( i \in [0, 1] \) and \( t = 0, 1, 2, ... \). Defining the real wage \( \tilde{W}_t = W_t/P_t \), the gross rate of price inflation \( \pi_t = P_t/P_{t-1} \), the system of equilibrium conditions becomes

1. \( Y_t \)

\[
Y_t = C_t + I_t + \left[ \phi_u (u_t - 1) + \frac{\phi_u^2}{2} (u_t - 1)^2 \right] K_{t-1}
\]

2. \( N_t \)

\[
N_t = \frac{\phi_V V_t + (1 - \phi_V) m_t}{N_t}
\]

3. \( m_t \)

\[
m_t = q_t V_t
\]

4. \( x_t \)

\[
x_t = \frac{m_t}{N_t}
\]

5. \( K_t \)

\[
K_t = u_t K_{t-1}
\]

6. \( \tilde{K}_t \)

\[
\tilde{K}_t = (1 - \delta) \tilde{K}_{t-1} + \mu_t \left[ 1 - \frac{\phi_f}{2} \left( \frac{I_t}{I_{t-1}} - g_t \right) \right] I_t
\]

7. \( \mu_t \)

\[
\ln \mu_t = \rho_\mu \ln \mu_{t-1} + \varepsilon_{\mu t}
\]

8. \( \varepsilon_{\mu t} \)

\[
\ln \varepsilon_{\mu t} = \rho_\varepsilon \ln \varepsilon_{\mu t-1} + \varepsilon_{\mu t}
\]
9. \( \Lambda_t \)
\[
\Lambda_t = \beta \varepsilon_{ut} R_t E_t \left( \frac{\Lambda_{t+1}}{\pi_{t+1}} \right)
\]

10. \( C_t \)
\[
\Lambda_t = \frac{1}{C_t - hC_{t-1}} - \beta h E_t \left( \frac{1}{C_{t+1} - hC_t} \right)
\]

11. \( \tilde{r}_t^K = \frac{r^K}{I_t} \)
\[
(\phi_{u1} - \phi_{u2}) + \phi_{u2} u_t = \tilde{r}_t^K
\]

12. \( I_t \)
\[
1 = v_t \mu_t \left[ 1 - \frac{\phi_I}{2} \left( \frac{I_t}{I_{t-1}} - gt \right)^2 - \phi_I \left( \frac{I_t}{I_{t-1}} - gt \right) \left( \frac{I_t}{I_{t-1}} \right) \right] + \beta E_t v_{t+1} \mu_{t+1} \frac{\Lambda_{t+1}}{\Lambda_t} \phi_I \left( \frac{I_{t+1}}{I_t} - gt \right) \left( \frac{I_{t+1}}{I_t} \right)^2
\]

13. \( v_t = \frac{\gamma_t}{\lambda_t} \)
\[
v_t = \beta E_t \left\{ \frac{\Lambda_{t+1}}{\Lambda_t} \left[ (1 - \delta) u_{t+1} + \tilde{r}_t^K u_{t+1} - a(u_{t+1}) \right] \right\}
\]

14. \( \theta_t \)
\[
\ln \theta_t = (1 - \rho_0) \ln \theta + \rho_0 \ln \theta_{t-1} + \varepsilon_{\theta_t}
\]

15. \( N_t \)
\[
N_t = \chi N_{t-1} + q_t V_t
\]

16. \( S_t \)
\[
S_t = 1 - \chi N_{t-1}
\]

17. \( U_t \)
\[
U_t = 1 - N_t
\]

18. \( \Theta_t = \frac{V_t}{S_t} \)
\[
\Theta_t = \frac{V_t}{S_t}
\]
19. \( q_t \)

\[
q_t = \zeta_t \left( \frac{S_t}{V_t} \right)^\sigma
\]

\[
q_t = \zeta_t \left( \frac{V_t}{S_t} \right)^{-\sigma}
\]

\[
q_t = \zeta_t \Theta_t^{-\sigma}
\]

20. \( s_t \)

\[
s_t = \zeta_t \left( \frac{V_t}{S_t} \right)^{1-\sigma}
\]

\[
s_t = \zeta_t \Theta_t^{1-\sigma}
\]

21. \( \zeta_t \)

\[
\ln \zeta_t = (1 - \rho_z) \ln \zeta + \rho_z \ln \zeta_{t-1} + \varepsilon_{zt}
\]

22. \( V_t \)

\[
\frac{\kappa Y_t}{m_t} = \xi_t (1 - \alpha) \frac{Y_t}{N_t} - \tilde{W}_t - \frac{\phi W}{2} \left( \frac{W_t}{z_{t-1}^{1-\xi} e^{-W_{t-1}}} - 1 \right)^2 Y_t
\]

\[+ \frac{\kappa}{N_t} \xi_t \left( 1 - \alpha \right) \frac{Y_{t+1}}{N_{t+1}} \frac{\Lambda_{t+1}}{\Lambda_t} \left( \frac{Y_{t+1}}{m_{t+1}} \right)
\]

23. \( u_t \)

\[Y_t = A_t^{1-\alpha} K_t \alpha N_t^{1-\alpha}\]

24. \( A_t \)

\[
z_t = \frac{A_t}{A_{t-1}}
\]

25. \( z_t = \frac{A_t}{A_{t-1}} \)

\[
\ln (z_t) = (1 - \rho_z) \ln (z) + \rho_z \ln (z_{t-1}) + \varepsilon_{zt}
\]

26. \( \xi_t \)

\[
\tilde{\xi}_t^K = \left( \alpha \frac{Y_t}{K_t} \right) \xi_t
\]

27. \( \pi_t \)

\[
\phi_p \left( \frac{\pi_t}{\pi_{t-1}^{1-\xi}} - 1 \right) \left( \frac{\pi_t}{\pi_{t-1}^{1-\xi}} \right) = (1 - \theta_t) + \theta_t \xi_t
\]

\[+ \beta \phi_p E_t \left[ \left( \frac{\Lambda_{t+1}}{\Lambda_t} \right) \left( \frac{Y_{t+1}}{Y_t} - 1 \right) \left( \frac{\pi_{t+1}}{\pi_t^{1-\xi}} \right) \right]
\]
28. $\tilde{b}_t = \frac{b_t}{\bar{r}}$

$\tilde{b}_t = \tau \bar{W}_{s.s, t}$

29. $\tilde{W}_t = \frac{W_t}{\bar{r}}$

$$
\tilde{W}_t = \left( \frac{\Omega_t}{1 + \Omega_t} \right) \left[ \xi_t (1 - \alpha) \frac{Y_t}{N_t} - \frac{\phi_W}{2} \left( \frac{W_t}{z \pi_{t-1} q^{1-\epsilon W_{t-1}}} - 1 \right)^2 Y_t + \frac{\kappa}{N_t} N_t^2 Y_t \right] \\
+ \left( \frac{\Omega_t}{1 + \Omega_t} \right) \left[ \beta \chi E_t \left( \frac{\Lambda_{t+1}}{\Lambda_t} \right) \kappa Y_t+1 \right] \\
+ \frac{1}{(1 + \Omega_t)} \left[ \tilde{b}_t - \beta \chi E_t \Omega_{t+1} (1 - s_{t+1}) \left( \frac{\Lambda_{t+1}}{\Lambda_t} \right) \kappa Y_t+1 \right]
$$

30. $\Omega_t$

$$
\Omega_t = \frac{W_t}{Y_t} + \phi_W \left( \frac{W_t}{z \pi_{t-1} q^{1-\epsilon W_{t-1}}} - 1 \right) \left( \frac{W_t}{z \pi_{t-1} q^{1-\epsilon W_{t-1}}} \right) - \beta \chi \phi_W E_t \left[ \frac{\Lambda_{t+1}}{\Lambda_t} \left( \frac{W_t+1}{z \pi_{t} q^{1-\epsilon W_{t}}} - 1 \right) \left( \frac{W_t+1}{z \pi_{t} q^{1-\epsilon W_{t}}} \right) \frac{Y_{t+1}}{Y_t} \right]
$$

31. $\eta_t$

$$
\ln \eta_t = (1 - \rho_{\eta}) \ln \eta + \rho_{\eta} \ln \eta_{t-1} + \varepsilon_{\eta t}
$$

32. $R_t$

$$
\ln \left( \frac{R_t}{R} \right) = \rho_r \ln \left( \frac{R_{t-1}}{R} \right) + (1 - \rho_r) \left\{ \rho \pi \ln \left[ \left( \frac{P_t/P_{t-4}}{\Pi} \right)^{1/4} \right] + \rho_g \ln \left[ \left( \frac{Y_t/Y_{t-4}}{G_y} \right)^{1/4} \right] \right\} + \ln \epsilon_{mpt}
$$

33. $\epsilon_{mpt}$

$$
\ln \epsilon_{mpt} = \rho_{mp} \ln \epsilon_{mpt-1} + \varepsilon_{mpt}
$$

34. $G_t$

$$
G_t = \left( 1 - \frac{1}{e_{gt}} \right) Y_t
$$

35. $\epsilon_{gt}$

$$
\ln \epsilon_{gt} = (1 - \rho_{g}) \ln \epsilon_{g} + \rho_{g} \ln \epsilon_{gt-1} + \varepsilon_{gt}
$$

36. $gy_t$ : Quarterly gross rate of output growth

$$
gy_t = Y_t/Y_{t-1}
$$

37. $gc_t$ : Quarterly gross rate of consumption growth

$$
 gc_t = C_t/C_{t-1}$$
38. \( g_i_t \): Quarterly gross rate of investment growth
\[
g_i_t = \frac{I_t}{I_{t-1}}
\]

39. \( g_w_t \): Quarterly gross rate of real wage growth
\[
g_w_t = \frac{\widetilde{W}_t}{\widetilde{W}_{t-1}}
\]

These 39 equations determine equilibrium values for the 39 variables \( Y_t, K_t, \bar{K}_t, u_t, C_t, A_t, R_t, G_t, I_t, v_t, \tilde{r}_t^K, \xi_t, N_t, S_t, U_t, V_t, \bar{V}_t, m_t, x_t, \Theta_t, q_t, s_t, \bar{W}_t, \bar{D}_t, \tilde{b}_t, \pi_t, \mu_t, \epsilon_{bt}, A_t, z_t, \zeta_t, \theta_t, \eta_t, \epsilon_{mpt}, \epsilon_{gt}, g_y_t, g_c_t, g_i_t, g_w_t \).

### 2.8 The stationary transformed economy

Output, consumption, investment, capital and the real wage share the stochastic trend induced by the unit root process of neutral technological progress. We first rewrite the model in terms of stationary variables, and then loglinearize this transformed model economy around its steady state. This approximate model can then be solved using standard methods. The following variables are stationary and need not to be transformed:

\[
\begin{align*}
  u_t, R_t, \tilde{r}_t^K, v_t = \frac{Y_t}{A_t}, \xi_t, N_t, S_t, U_t, V_t, \bar{V}_t, m_t, x_t, q_t, s_t, \pi_t = \frac{\bar{D}_t}{\bar{D}_{t-1}}, \mu_t, a_t, z_t, \zeta_t, \theta_t, \eta_t, \epsilon_{mpt}, \epsilon_{gt} \text{ and } \bar{D}_t.
\end{align*}
\]

We define the transformed variables \( y_t = Y_t/A_t, k_t = K_t/A_t, \bar{k}_t = \bar{K}_t/A_t, c_t = C_t/A_t, \lambda_t = A_t A_t, i_t = I_t/A_t, \tilde{w}_t = \widetilde{W}_t/A_t, \tilde{b}_t = \tilde{b}_t/A_t, y_t = G_t/A_t \). The stationaryized economy contains only 38 equations in 38 variables because the level of the non-stationary productivity shock \( A_t \) is not included.

1. \( y_t = Y_t/A_t \)
\[
y_t = \frac{1}{\epsilon_{gt}} - \frac{\kappa_t}{2} \frac{N_t}{2} - \frac{\phi_p}{2} \left( \frac{\pi_t}{\pi_{t-1}^1} - 1 \right)^2 - \frac{\phi_{w t}}{2} \left( \frac{z_t \pi_t \tilde{w}_t}{z_{t-1} \pi_t^1 \tilde{w}_{t-1}} - 1 \right)^2 N_t \frac{1}{z_t}
\]

2. \( \bar{k}_t \)
\[
\bar{k}_t = \frac{\varphi_V V_t + (1 - \varphi_V) m_t}{N_t}
\]

3. \( m_t \)
\[
m_t = q_t V_t
\]

4. \( x_t \)
\[
x_t = \frac{m_t}{N_t}
\]

5. \( k_t = K_t/A_t \)
\[
k_t = u_t k_{t-1} \frac{1}{z_t}
\]
6. $\tilde{K}_t = \frac{K_t}{A_t}$

$$\tilde{K}_t = (1 - \delta) K_{t-1} \frac{1}{z_t} + \mu_t \left[ 1 - \frac{\phi_I}{2} \left( \frac{i_t}{i_{t-1}} z_t - g_I \right)^2 \right] i_t$$

7. $\mu_t$

$$\ln \mu_t = \rho_\mu \ln \mu_{t-1} + \varepsilon_\mu$$

8. $\varepsilon_{bt}$

$$\ln \varepsilon_{bt} = \rho_\varepsilon \ln \varepsilon_{bt-1} + \varepsilon_{bt}$$

9. $\lambda_t = A_t \Lambda_t$

$$\lambda_t = \beta \varepsilon_{bt} T_t E_t \left( \frac{\lambda_{t+1}}{\pi_{t+1}} \frac{1}{z_{t+1}} \right)$$

10. $c_t = C_t / A_t$

$$\lambda_t = \frac{z_t}{z_t c_t - hc_{t-1}} - \beta h E_t \left( \frac{1}{c_{t+1}} \frac{1}{z_{t+1}} - hc_t \right)$$

11. $\tilde{r}_t^K = \frac{r_t^K}{\tilde{E}_t}$

$$(\phi_{u1} - \phi_{u2}) + \phi_{u2} u_t = \tilde{r}_t^K$$

12. $i_t = I_t / A_t$

$$1 = v_t \mu_t \left[ 1 - \frac{\phi_I}{2} \left( \frac{i_t}{i_{t-1}} z_t - g_I \right)^2 \right] - \phi_I \left( \frac{i_t}{i_{t-1}} z_t - g_I \right) \left( \frac{i_t}{i_{t-1}} z_t \right)$$

$$+ \beta E_t v_{t+1} \mu_{t+1} \lambda_t \lambda_{t+1} \frac{1}{\pi_{t+1}} \phi_I \left( \frac{i_{t+1}}{i_t} z_{t+1} - g_I \right) \left( \frac{i_{t+1}}{i_t} z_{t+1} \right)^2$$

13. $v_t = \frac{\lambda_{t+1}}{\lambda_t}$

$$v_t = \beta E_t \left\{ \frac{\lambda_{t+1}}{\lambda_t} \frac{1}{z_{t+1}} \left[ (1 - \delta) v_{t+1} + \tilde{r}_t^K u_{t+1} - \phi_{u1} (u_{t+1} - 1) - \frac{\phi_{u2}}{2} (u_{t+1} - 1)^2 \right] \right\}$$

14. $u_t$

$$y_t = k_t^\alpha N_t^{1-\alpha}$$

15. $z_t = \frac{A_t}{A_{t-1}}$

$$\ln z_t = (1 - \rho_z) \ln z + \rho_z \ln z_{t-1} + \varepsilon_{zt}$$
16. $\xi_t \equiv \frac{\eta_t}{\pi_t}$

$$\tilde{r}_t^K = \alpha \frac{y_t}{\pi_t} \xi_t$$

17. $N_t$

$$N_t = \chi \bar{N}_{t-1} + q_t \bar{V}_t$$

18. $S_t$

$$S_t = 1 - \chi \bar{N}_{t-1}$$

19. $U_t$

$$U_t = 1 - N_t$$

20. $\Theta_t = \frac{V_t}{S_t}$

$$\Theta_t = \frac{V_t}{S_t}$$

21. $q_t$

$$q_t = \zeta_t \Theta_t^{-\sigma}$$

22. $s_t$

$$s_t = \zeta_t \Theta_t^{1-\sigma}$$

23. $\zeta_t$

$$\ln \zeta_t = (1 - \rho_{\zeta}) \ln \zeta + \rho_{\zeta} \ln \zeta_{t-1} + \varepsilon_{\zeta_t}$$

24. $V_t$

$$\frac{\kappa N_t^2 y_t}{m_t} = \xi_t (1 - \alpha) y_t \frac{\bar{N}_t}{N_t} - \bar{w}_t - \frac{\phi W}{2} \left( \frac{\pi_t \pi_{t-1} \bar{w}_{t-1}}{z \pi_{t-1} \pi_{t-1} - 1} - 1 \right) y_t + \frac{\kappa N_t^2 y_t}{N_t} + \beta \chi \frac{\lambda_{t+1} \kappa N_t^2 y_{t+1}}{\lambda_t m_{t+1}}$$

25. $\theta_t$

$$\ln \theta_t = (1 - \rho_{\theta}) \ln \theta + \rho_{\theta} \ln \theta_{t-1} + \varepsilon_{\theta_t}$$

26. $\pi_t = \frac{P_t}{\pi_{t-1}}$

$$0 = (1 - \theta_t) + \theta_t \xi_t - \phi_P \left( \frac{\pi_t}{\pi_{t-1} \pi_{t-1} - 1} \right) \left( \frac{\pi_t}{\pi_{t-1} \pi_{t-1} - 1} \right)$$

$$+ \beta \phi_P E_t \left[ \frac{\lambda_{t+1}}{\lambda_t} \left( \frac{\pi_{t+1}}{\pi_t \pi_{t-1} - 1} \right) \left( \frac{\pi_{t+1}}{\pi_t \pi_{t-1} - 1} \right) \frac{y_{t+1}}{y_t} \right]$$
27. \( \ddot{b}_t = \frac{\ddot{b}_t}{A_t} \)

\[ \ddot{b}_t = \ddot{b} = \tau \ddot{w} \]

28. \( \ddot{w}_t = \frac{\ddot{W}_t}{A_t} \)

\[
\ddot{w}_t = \left[ \frac{\Omega_t}{1 + \Omega_t} \right] \left[ \xi_t (1 - \alpha) \frac{y_t}{N_t} - \frac{\phi W}{2} \left( \frac{z_t \pi t \ddot{w}_t}{z \pi t - 1 \pi^{1 - \varphi} \ddot{w}_{t-1}} - 1 \right)^2 \right] y_t + \frac{\kappa \mathbb{N}_t^2 y_t}{N_t} + \beta \chi E_t \left( \frac{\lambda_{t+1}}{\lambda_t} \right) \frac{\kappa \mathbb{N}_t^2 y_{t+1}}{m_{t+1}} \\
+ \frac{1}{1 + \Omega_t} \left[ \ddot{b} - \beta \chi E_t \Omega_{t+1} (1 - s_{t+1}) \frac{\lambda_{t+1} \kappa \mathbb{N}_t^2 y_{t+1}}{m_{t+1}} \right]
\]

29. \( \Omega_t \)

\[ \Omega_t = \left[ \left( \frac{\eta_t}{1 - \eta_t} \right) \ddot{w}_t \right] / \left( \frac{\ddot{w}_t}{y_t} \right) \]

\[ - \beta \chi \phi W E_t \left[ \frac{\lambda_{t+1}}{\lambda_t} \frac{\lambda_{t+1} \kappa \mathbb{N}_t^2 y_{t+1}}{m_{t+1}} \right] \]

30. \( \eta_t \)

\[ \ln \eta_t = (1 - \rho_{\eta}) \ln \eta + \rho_{\eta} \ln \eta_{t-1} + \varepsilon_{\eta_t} \]

31. \( R_t \)

\[ \ln \left( \frac{R_t}{R} \right) = \rho_r \ln \left( \frac{R_{t-1}}{R} \right) + (1 - \rho_r) \rho_g \ln \left[ \frac{(\pi_t \pi_{t-1} \pi_{t-2} \pi_{t-3})^{1/4}}{\pi} \right] \]

\[ + (1 - \rho_r) \rho_g \ln \left[ \frac{(g_y g_y y t - 1 g_y y t - 2 g_y y t - 3)^{1/4}}{g} \right] + \ln \epsilon_{mpt} \]

32. \( \epsilon_{mpt} \)

\[ \ln \epsilon_{mpt} = \rho_{mpt} \ln \epsilon_{mpt-1} + \varepsilon_{mpt} \]

33. \( g_t = G_t / A_t \)

\[ g_t = \left( 1 - \frac{1}{\epsilon_{g t}} \right) y_t \]

34. \( \epsilon_{g t} \)

\[ \ln \epsilon_{g t} = (1 - \rho_g) \ln \epsilon_g + \rho_g \ln \epsilon_{g t-1} + \varepsilon_{g t} \]

35. \( g_y t = Y_t / Y_{t-1} \)

\[ g_y t = \frac{y_t}{y_{t-1}} \]

36. \( g_c t = C_t / C_{t-1} \)

\[ g_c t = \left( \frac{c_t}{c_{t-1}} \right) \]

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2.9 The steady state of the transformed economy

In the absence of shocks, the economy converges to a steady-state growth path in which all stationary variables are constant: for all \( t; y_t = y; k_t = k; \lambda_t = \lambda; v_t = v; \xi_t = \xi; c_t = c; \tilde{K}_t = \tilde{K}; \)
\( i_t = i; g_t = g; N_t = N; S_t = S; U_t = U; V_t = V; \beta = \beta; m_t = m; x_t = x; q_t = q; s_t = s; \Pi_t = \Pi; \tilde{w}_t = \tilde{w}; \)
\( b = b; R_t = R; \pi_t = \pi; \mu_t = \mu = 1; \epsilon_{bt} = \epsilon_b = 1; z_t = z; \zeta_t = \zeta; \theta_t = \theta; \eta_t = \eta; \epsilon_{gt} = \epsilon_g; \epsilon_{rt} = \epsilon_r; \)
\( g_{yt} = g_{ct} = g_{it} = g_{At} = z. \) Notice that the steady-state values \( \mu, u \) and \( \epsilon_b \) are normalized to 1.

1. \( \mu_t \)
   \[ \mu = 1 \]

2. \( \epsilon_{bt} \)
   \[ \epsilon_b = 1 \]

3. \( u_t \)
   \[ u = 1 \]

4. \( z_t \)
   \[ z : \text{calibrated at sample mean of gross quarterly growth rate of per-capita real output} \]

5. \( g_{yt} \)
   \[ g_y = z \]

6. \( g_{ct} \)
   \[ g_c = z \]

7. \( g_{it} \)
   \[ g_i = z \]

8. \( gw_t \)
   \[ gw = z \]
9. \( g_t \)
\[
\frac{g}{y} = \left(1 - \frac{1}{\epsilon_g}\right) = 0.20 \text{ (calibrated)}
\]

10. \( e_{gt} \)
\[
\frac{1}{\epsilon_g} - \frac{\kappa}{2} \mathbb{N}_t^2 = \frac{c + i}{y}
\]

11. \( \mathbb{N}_t \)
\[
\mathbb{N} = \phi_N V + (1 - \phi_Y) \frac{m}{N}
\]

12. \( m_t \)
\[
m = qV
\]

13. \( x_t \)
\[
x = \frac{m}{N}
\]

14. \( k_t \)
\[
z k = \bar{k}
\]

15. \( \bar{k}_t \)
\[
(z - 1 + \delta) \bar{k} = zi
\]

16. \( \lambda_t \)
\[
\beta = \frac{\pi z}{R}
\]

17. \( c_t \)
\[
\lambda c = \frac{z - \beta h}{z - h}
\]

18. \( \bar{t}_t^K \)
\[
\phi_{a1} = \bar{t}_t^K
\]

19. \( i_t \)
\[
1 = v
\]
20. $v_t$
\[ \frac{z}{\beta} = 1 - \delta + \tilde{r}^K \]

21. $N_t$
\[ \rho N = qV \quad \text{where} \quad \rho \equiv 1 - \chi \]

22. $S_t$
\[ S = 1 - \chi N \]

23. $U_t$

\[ U : \text{calibrated at sample mean of unemployment rate} \]

24. $\Theta_t = \frac{V}{S_t}$
\[ \Theta = \frac{V}{S} \]

25. $q_t$
\[ q = \zeta \Theta^{-\sigma} := 0.7 \quad \text{(calibrated. just a normalization)} \]

26. $s_t$
\[ s = \zeta \Theta^{1-\sigma} \]

27. $\zeta_t$
\[ \zeta : \text{backed out from the steady state condition} \quad \zeta = q \left( \frac{V}{S} \right)^{\sigma} \]

28. $y_t$
\[ y = k^\alpha N^{1-\alpha} \]

29. $\xi_t$
\[ \tilde{r}^K = \alpha \frac{y}{k} \xi \]

30. $V_t$
\[ \frac{(1 - \beta) \chi}{\rho} \kappa N^2 = \xi (1 - \alpha) - \frac{\tilde{w} N}{y} \]

31. $\theta_t$
\[ \xi = \frac{\theta - 1}{\theta} \]
32. $\pi_t$

$\pi$: calibrated at sample mean of gross quarterly growth rate GDP deflator

33. $\tilde{b}_t$

$\tilde{b} = \tau \tilde{w}$

34. $\tilde{w}_t$

$$\tilde{w} = \eta \left[ (1 - \alpha) \frac{\xi y}{N} + \left( \frac{1}{N} + \chi \frac{\beta}{m} \right) \kappa N^2 y \right] + (1 - \eta) \tilde{b}$$

$$\Leftrightarrow \frac{1}{\eta} - \frac{1 - \eta}{\eta^2} \tau \tilde{w} N \frac{y}{y} = \xi (1 - \alpha) + \left( 1 + \beta \chi \frac{\beta}{\rho} \right) \kappa N^2$$

35. $\Omega_t$

$$\Omega = \frac{\eta}{1 - \eta}$$

36. $\eta_t$

$\eta$: backed out from steady state conditions (see Table 4 below)

37. $\epsilon_{rt}$

$$\epsilon_{mp} = 1$$

38. $R_t$

$R$: calibrated at sample mean of gross quarterly nominal rate of interest

2.10 The loglinear model with rescaled shocks

Two disturbances are normalized prior to estimation: the price-markup shock $\hat{\theta}_t$ and the wage bargaining shock $\hat{\eta}_t$. Rescaling these two shocks only affects the New Keynesian Phillips Curve and the equation for the evolution of the effective bargaining power.

$$\hat{\theta}_t^* = \left[ \frac{1}{(1 + \beta \xi) \phi_p} \right] \hat{\theta}_t$$

$$\hat{\eta}_t^* = \rho_{\theta^*} \hat{\eta}_{t-1} - \varepsilon_{\theta^*}$$

$$\rho_{\theta^*} = \rho_{\theta}$$

$$\varepsilon_{\theta^*} \sim i.i.d.N \left( 0, \sigma_{\theta^*}^2 \right)$$

$$\sigma_{\theta^*} = \left[ \frac{1}{(1 + \beta \xi) \phi_p} \right] \sigma_{\theta}$$
\[
\begin{align*}
\hat{\eta}_t &= \left( \frac{1}{1-\eta} \right) \hat{n}_t \\
\hat{\eta}_t^* &= \rho_{\eta} \hat{\eta}_{t-1} + \varepsilon_{\eta,t} \\
\rho_{\eta} &= \rho \\
\varepsilon_{\eta,t} &\sim \text{i.i.d.} N \left( 0, \sigma_{\eta}^2 \right) \\
\sigma_{\eta} &= \left( \frac{1}{1-\eta} \right) \sigma_{\eta} \\
1. \ y_t & \\
\frac{c + i}{y} \hat{y}_t &= \frac{c}{y} \hat{c}_t + \frac{i}{y} \hat{t}_t + \phi_{u1} \frac{k}{y} \hat{u}_t + \frac{1}{\varepsilon_{\eta,t}} + \kappa \hat{\varepsilon}_t \\
2. \ k_t & \\
\hat{k}_t &= \hat{u}_t + \hat{k}_{t-1} - \hat{z}_t \\
3. \ \bar{k}_t & \\
\hat{z}_t = (1 - \delta) \hat{k}_{t-1} - (1 - \delta) \hat{z}_t + (z - 1 + \delta) \hat{\mu}_t + (z - 1 + \delta) \hat{\mu}_t \\
4. \ \lambda_t & \\
\hat{\lambda}_t &= \varepsilon_{bl} + \hat{R}_t + \hat{\lambda}_{t+1} - \hat{\pi}_{t+1} - \hat{z}_{t+1} \\
5. \ c_t & \\
\hat{\lambda}_t &= \frac{\beta h z}{(z - \beta h)(z - h)} \hat{c}_{t+1} - \frac{z^2 + \beta h^2}{(z - \beta h)(z - h)} \hat{c}_t + \frac{h z}{(z - \beta h)(z - h)} \hat{c}_{t-1} \\
6. \ \tilde{r}_t^K & \\
\hat{r}_t^K &= \left( \frac{\phi_{u2}}{\phi_{u1}} \right) \hat{u}_t \\
7. \ i_t & \\
\hat{v}_t &= \left[ (1 + \beta) (\phi I z^2) \right] \hat{u}_t + (\phi I z^2) \hat{z}_t - (\phi I z^2) \hat{z}_{t-1} - \hat{\mu}_t - (\beta \phi I z^2) \hat{\pi}_{t+1} + (\beta \phi I z^2) \hat{z}_{t+1} \\
8. \ v_t & \\
\hat{v}_t &= \hat{\lambda}_{t+1} - \hat{\lambda}_t - \hat{z}_{t+1} + \left[ (1 - \delta) \beta z^{-1} \right] \hat{v}_{t+1} + (\beta z^{-1} \tilde{r}_K) \hat{r}_{t+1} \\
9. \ u_t & \\
\hat{y}_t &= \alpha \hat{k}_t + (1 - \alpha) \hat{N}_t \\
\end{align*}
\]
10. $\xi_t$

\[ \hat{\xi}_t = \hat{r}_t^K - \hat{y}_t + \hat{k}_t \]

11. $\pi_t$

\[ \hat{\pi}_t = \left( \frac{s}{1 + \beta_s} \right) \hat{\pi}_{t-1} + \left( \frac{\beta}{1 + \beta_s} \right) \hat{\pi}_{t+1} + \left( \frac{1}{1 + \beta_s} \right) \left( \frac{\theta - 1}{\phi_p} \right) \hat{\xi}_t - \hat{\theta}_t^* \]

12. $N_t$

\[ \hat{N}_t = \chi \hat{N}_{t-1} + (1 - \chi) \hat{q}_t + (1 - \chi) \hat{V}_t \]

13. $U_t$

\[ \hat{U}_t = - \left( \frac{N}{1 - N} \right) \hat{N}_t \]

14. $\Theta_t$

\[ \hat{\Theta}_t = \hat{V}_t + \left( \frac{\chi N}{S} \right) \hat{N}_{t-1} \]

15. $q_t$

\[ \hat{q}_t = \hat{\zeta}_t - \sigma \hat{\Theta}_t \]

16. $s_t$

\[ \hat{s}_t = \hat{\zeta}_t + (1 - \sigma) \hat{\Theta}_t \]

17. $R_t$

\[ \hat{R}_t = \left[ \frac{\phi_y V}{\phi_y V + (1 - \phi_y) m} \right] \hat{V}_t + \left[ \frac{(1 - \phi_y) m}{\phi_y V + (1 - \phi_y) m} \right] \hat{m}_t - \hat{N}_t \]

18. $m_t$

\[ \hat{m}_t = \hat{q}_t + \hat{V}_t \]

19. $x_t$

\[ \hat{x}_t = \hat{m}_t - \hat{N}_t \]

20. $V_t$

\[ 2 \frac{\chi}{\rho} \kappa N^2 \hat{R}_t = \frac{\kappa N^2}{\rho} \hat{x}_t + (1 - \alpha) \hat{\xi}_t + \left( \frac{\bar{w} N}{y} - \frac{\beta \chi}{\rho} \kappa N^2 \right) \left( \hat{y}_t - \hat{N}_t \right) - \frac{\bar{w} N}{y} \hat{w}_t + \frac{\beta \chi}{\rho} \kappa N^2 \left( \hat{\lambda}_{t+1} - \hat{\lambda}_t + \hat{y}_{t+1} - \hat{N}_{t+1} + 2 \hat{N}_{t+1} - \hat{x}_{t+1} \right) \]
21. \[\frac{1}{\eta} R_t = (1 - \alpha) \zeta \tilde{\xi}_t + \left[ (1 - \alpha) \zeta + \kappa \eta^2 \right] \left( \tilde{y}_t - \tilde{N}_t \right) + 2\kappa \eta^2 \tilde{N}_t + \beta \chi \frac{s}{\rho} \kappa \eta^2 \left( \tilde{\gamma}_{t+1} + \tilde{\lambda}_{t+1} - \tilde{\lambda}_t + 2\tilde{N}_{t+1} + \tilde{y}_{t+1} - \tilde{m}_{t+1} \right) \]
\[\quad \quad \quad - \left[ \frac{\tilde{w}_N}{\eta} - (1 - \alpha) \zeta - \left( 1 + \frac{\beta \chi}{\rho} \right) \kappa \eta^2 \right] \tilde{N}_t - \beta \chi \frac{(1 - s)}{\rho} \kappa \eta^2 \tilde{N}_{t+1}\]

22. \[\hat{\Omega}_t = \hat{\eta}_t^* + (\beta \chi \phi_W \frac{y}{w}) \hat{\tilde{\gamma}}_{t+1} + (\beta \chi \phi_W \frac{y}{w}) \hat{\tilde{\pi}}_{t+1} + \left( \beta \chi \phi_W \frac{y}{w} \right) \hat{\tilde{w}}_{t+1} - \left[ \left( \phi_W \frac{y}{w} \right) (1 + \beta \chi) \right] \hat{\tilde{w}}_t - \left[ (\phi_W \frac{y}{w}) (1 + \beta \chi \phi) \right] \hat{\tilde{\pi}}_t + \left( \phi_W \frac{y}{w} \right) \hat{\tilde{w}}_{t-1} + \left( \phi_W \frac{y}{w} \right) \hat{\tilde{w}}_{t-1}\]

23. \[\hat{R}_t = \rho_r \hat{R}_{t-1} + \frac{1 - \rho_r}{4} \rho_y (\hat{\tilde{\pi}}_t + \hat{\tilde{\pi}}_{t-1} + \hat{\tilde{\pi}}_{t-2} + \hat{\tilde{\pi}}_{t-3}) + \frac{1 - \rho_r}{4} \rho_y (\bar{gy}_t + \bar{gy}_{t-1} + \bar{gy}_{t-2} + \bar{gy}_{t-3}) + \hat{\epsilon}_{mpt}\]

24. \[gy_t = Y_t / Y_{t-1}\]
\[\bar{gy}_t = \bar{y}_t - \bar{y}_{t-1} + \hat{\tilde{z}}_t\]

25. \[gc_t = C_t / C_{t-1}\]
\[\hat{gc}_t = \hat{c}_t - \hat{c}_{t-1} + \hat{\tilde{z}}_t\]

26. \[gi_t = I_t / I_{t-1}\]
\[\hat{gi}_t = \hat{i}_t - \hat{i}_{t-1} + \hat{\tilde{z}}_t\]

27. \[gw_t = \bar{W}_t / \bar{W}_{t-1}\]
\[\bar{g}w_t = \hat{\tilde{w}}_t - \hat{\tilde{w}}_{t-1} + \hat{\tilde{z}}_t\]

28. \[\mu_t\]
\[\hat{\mu}_t = \rho_\mu \hat{\mu}_{t-1} + \hat{\epsilon}_{\mu t}\]

29. \[\epsilon_{\mu t}\]
\[\hat{\epsilon}_{\mu t} = \rho_\epsilon \hat{\epsilon}_{\mu t-1} + \hat{\epsilon}_{\mu t}\]

30. \[z_t\]
\[\hat{\tilde{z}}_t = \rho_z \hat{\tilde{z}}_{t-1} + \hat{\epsilon}_{zt}\]
31. $\zeta_t$

$$\hat{\zeta}_t = \rho_\zeta \hat{\zeta}_{t-1} + \varepsilon_{\zeta t}$$

32. $\theta_t$

$$\hat{\theta}_t = \rho_\theta \hat{\theta}_{t-1} + \varepsilon_{\theta t}$$

33. $\eta_t$

$$\hat{\eta}_t = \rho_\eta \hat{\eta}_{t-1} + \varepsilon_{\eta t}$$

34. $\epsilon_{gt}$

$$\hat{\epsilon}_{gt} = \rho_{\epsilon_g} \hat{\epsilon}_{gt-1} + \varepsilon_{\epsilon gt}$$

35. $\epsilon_{mt}$

$$\hat{\epsilon}_{mpt} = \rho_{\epsilon_{mpt}} \hat{\epsilon}_{mpt-1} + \varepsilon_{\epsilon mpt}$$

2.11 Natural equilibrium: no nominal rigidities, no markup shocks and no bargaining power shocks

We compute the natural equilibrium by setting equal to zero the two parameters $\phi_p$ and $\phi_W$ that govern the degree of nominal rigidities in prices and wages respectively and by turning off the price-markup shock $\theta_t$ and the bargaining-power shock $\eta_t$.

1. $\epsilon_t^p$

$$\frac{c + i \epsilon_t^p}{y} = \frac{c^p}{y} + \frac{i}{y} \hat{\epsilon}_t^p + \phi_{u1} \frac{k}{y} \hat{u}_t^p + \kappa \lambda^2 \hat{\lambda}_t^p + 1 \hat{\epsilon}_{gt}$$

2. $k_t^p$

$$\hat{k}_t^p = \hat{u}_t^p + \hat{\kappa}_{t-1} - \hat{z}_t$$

3. $\tilde{k}_t^p$

$$\tilde{z}_t^p = (1 - \delta) \tilde{k}_{t-1}^p - (1 - \delta) \hat{z}_t + (z - 1 + \delta) \hat{\mu}_t + (z - 1 + \delta) \hat{\gamma}_t^p$$

4. $\tilde{R}_t^p$

$$\tilde{\lambda}_t^p = \tilde{c}_t + \tilde{\kappa}_t^p + \tilde{\lambda}_{t+1}^p - \tilde{z}_{t+1}$$
5. $\lambda_i^p$

$$\hat{\lambda}_t^p = \frac{\beta h z}{(z - \beta h)(z - h)} \hat{\lambda}_{t+1}^p - \frac{z^2 + \beta h^2}{(z - \beta h)(z - h)} \hat{\lambda}_t^p + \frac{h z}{(z - \beta h)(z - h)} \hat{\lambda}_{t-1}^p$$

$$+ \frac{h z}{(z - \beta h)(z - h)} \hat{z}_{t+1} - \frac{h z}{(z - \beta h)(z - h)} \hat{z}_t$$

6. $u_t^p$

$$\hat{z}_{t+1}^K p = \left( \frac{\phi_{u2}}{\phi_{u1}} \right) \hat{u}_t^p$$

7. $i_t^p$

$$\hat{v}_t^p = [(1 + \beta) (\phi_I z^2)] \hat{v}_t^p + (\phi_I z^2) \hat{z}_t - (\phi_I z^2) \hat{v}_{t-1}^p - \hat{\mu}_t - (\beta \phi_I z^2) \hat{v}_{t+1}^p - (\beta \phi_I z^2) \hat{z}_{t+1}$$

8. $v_t^p$

$$\hat{v}_t^p = \hat{\lambda}_{t+1}^p - \hat{\lambda}_t^p - \hat{z}_{t+1} + [(1 - \delta) \beta z^{-1}] \hat{v}_{t+1}^p + (\beta z^{-1} \alpha^K) \hat{r}_{t+1}$$

9. $y_t^p$

$$\hat{y}_t^p = \alpha \hat{y}_t^p + (1 - \alpha) \hat{N}_t^p$$

10. $\tilde{r}_t^p$

$$\hat{r}_t^p = \hat{z}_t^p - \hat{y}_t^p$$

11. $N_t^p$

$$\hat{N}_t^p = \chi \hat{N}_{t-1}^p + (1 - \chi) \hat{q}_t^p + (1 - \chi) \hat{V}_t^p$$

12. $U_t^p$

$$\hat{U}_t^p = -\left( \frac{N}{1-N} \right) \hat{N}_t^p$$

13. $\Theta_t^p$

$$\hat{\Theta}_t^p = \hat{v}_t^p + \left( \frac{\chi N}{S} \right) \hat{N}_{t-1}^p$$

14. $q_t^p$

$$\hat{q}_t^p = \hat{\zeta}_t - \sigma \hat{\Theta}_t^p$$

15. $s_t^p$

$$\hat{s}_t^p = \hat{\zeta}_t + (1 - \sigma) \hat{\Theta}_t^p$$
16. $N_t^p$

$$\hat{N}_t^p = \left[ \frac{\phi V}{\phi V + (1 - \phi V) m} \right] \hat{V}_t^p + \left[ \frac{(1 - \phi V) m}{\phi V + (1 - \phi V) m} \right] \hat{m}_t^p - \hat{N}_t^p$$

17. $m_t^p$

$$\hat{m}_t^p = \hat{a}_t^p + \hat{V}_t^p$$

18. $x_t^p$

$$\hat{x}_t^p = \hat{m}_t^p - \hat{N}_t^p$$

19. $V_t^p$

$$2 \frac{\chi}{\rho} \kappa N^2 \hat{N}_t^p = \frac{\kappa N^2}{\rho} \hat{X}_t^p + \left( \frac{\tilde{w} N}{y} - \frac{\beta \chi}{\rho} \kappa N^2 \right) \left( \hat{V}_t^p - \hat{N}_t^p \right) - \frac{\tilde{w} N}{y} \hat{w}_t^p$$

$$+ \frac{\beta \chi}{\rho} \kappa N^2 \left( \hat{X}_{t+1}^p - \hat{X}_t^p + \hat{y}_{t+1}^p - \hat{N}_{t+1}^p + 2 \hat{N}_{t+1}^p - \hat{x}_{t+1}^p \right)$$

20. $\tilde{w}_t^p$

$$\frac{1}{\eta} \frac{\tilde{w} N \hat{w}_t^p}{y} = \left[ (1 - \alpha) \xi + \kappa N^2 \right] \left( \hat{V}_t^p - \hat{N}_t^p \right) + 2 \kappa N^2 \hat{N}_t^p$$

$$+ \beta \chi \frac{\kappa N^2}{\rho} \left( \hat{X}_{t+1}^p + \hat{X}_t^p - \hat{x}_t^p + 2 \hat{N}_{t+1}^p + \hat{y}_{t+1}^p - \hat{N}_{t+1}^p \right)$$
3 Empirical analysis

3.1 Calibrated parameters

We calibrate 13 parameters. The steady-state values of output growth, inflation, the interest rate and the unemployment rate are set equal to their respective sample average over the period 1957Q1-2008Q3 in the baseline estimation (or, in the sensitivity analysis: 1957Q1-2013Q2 and 1985Q1-2008Q3). The value for the elasticity of the matching function with respect to unemployment is based on the recent estimates obtained by Barnichon and Figura (2014), Justiniano and Michelacci (2011), Lubik (2013), Shimer (2005) and Sedlacek (2014). The calibration of the job destruction rate is based on Yashiv (2006). The calibration of the replacement rate is a conservative value advocated by Shimer (2005). These choices avoid indeterminacy issues that are widespread in this kind of model, as shown by Kurozumi and Van Zandweghe (2010) among others. In preliminary estimation rounds, the estimate of the parameter governing the degree of indexation to past inflation was systematically driven towards zero. This phenomenon is consistent with the findings reported by Ireland (2007). It is also in line with the microevidence on price-setting behavior. Hence we calibrate that parameter to 0.01. The quarterly depreciation rate is set equal to 0.025. The capital share of output is calibrated at 0.33. The elasticity of substitution between intermediate goods is set equal to 6, implying a steady-state markup of 20 percent as in Rotemberg and Woodford (1995). The vacancy-filling rate is set equal to 0.70, which is just a normalization. The steady-state government spending/output ratio is set equal to 0.20.

Table 1: Calibrated parameters

<table>
<thead>
<tr>
<th>Parameter</th>
<th>$57Q1 – 08Q3$</th>
<th>$57Q1 – 13Q2$</th>
<th>$85Q1 – 08Q3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Capital depreciation rate</td>
<td>$\delta$</td>
<td>0.0250</td>
<td>0.0250</td>
</tr>
<tr>
<td>Capital share</td>
<td>$\alpha$</td>
<td>0.33</td>
<td>0.33</td>
</tr>
<tr>
<td>Elasticity of substitution btw goods</td>
<td>$\theta$</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>Backward-looking price setting</td>
<td>$\zeta$</td>
<td>0.01</td>
<td>0.01</td>
</tr>
<tr>
<td>Replacement rate</td>
<td>$\tau$</td>
<td>0.4</td>
<td>0.4</td>
</tr>
<tr>
<td>Job destruction rate</td>
<td>$\rho$</td>
<td>0.085</td>
<td>0.085</td>
</tr>
<tr>
<td>Elasticity of matches to unemp.</td>
<td>$\sigma$</td>
<td>0.65</td>
<td>0.65</td>
</tr>
<tr>
<td>Probability to fill a vacancy within a quarter</td>
<td>$q$</td>
<td>0.7000</td>
<td>0.7000</td>
</tr>
<tr>
<td>Exogenous spending/output ratio</td>
<td>$g/y$</td>
<td>0.2000</td>
<td>0.2000</td>
</tr>
<tr>
<td>Unemployment rate</td>
<td>$U$</td>
<td>0.0578</td>
<td>0.0602</td>
</tr>
<tr>
<td>Quarterly gross growth rate</td>
<td>$z$</td>
<td>1.0039</td>
<td>1.0038</td>
</tr>
<tr>
<td>Quarterly gross inflation rate</td>
<td>$\pi$</td>
<td>1.0088</td>
<td>1.0083</td>
</tr>
<tr>
<td>Quarterly gross nominal interest rate</td>
<td>$R$</td>
<td>1.0139</td>
<td>1.0128</td>
</tr>
</tbody>
</table>

3.2 Bayesian estimation

Our priors are standard (Smets and Wouters 2007; Gertler, Sala and Trigari 2008). We normalize the price-markup shock and the wage-markup shock, so that these enter with a unit coefficient in the model’s equations. Such procedure facilitates the identification of the standard deviations of these two disturbances.
We use the random walk Metropolis-Hasting algorithm to generate 500,000 draws from the posterior distribution. The algorithm is tuned to achieve an acceptance ratio between 25 and 30 percent. We discard the first 250,000 draws. Tables 2 and 3 summarize the priors.

### Table 2: Priors of structural parameters

<table>
<thead>
<tr>
<th>Priors</th>
<th>Priors</th>
</tr>
</thead>
<tbody>
<tr>
<td>Weight of pre-match cost in total hiring cost $\phi_V$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>Hiring cost/output ratio (rescaled) $\frac{\sigma^2}{\beta^2} \times 1000$</td>
<td>IGamma (5,1)</td>
</tr>
<tr>
<td>Habit persistence in consump. $h$</td>
<td>Beta (0.7,0.1)</td>
</tr>
<tr>
<td>Investment adjustment cost $\phi_I$</td>
<td>IGamma (5,1)</td>
</tr>
<tr>
<td>Capital utilization cost $\phi_u$</td>
<td>IGamma (0.5,0.1)</td>
</tr>
<tr>
<td>Price adjustment cost $\phi_p$</td>
<td>IGamma (60,10)</td>
</tr>
<tr>
<td>Wage adjustment cost $\phi_w$</td>
<td>IGamma (150,25)</td>
</tr>
<tr>
<td>Backward-looking wage setting $g$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>Interest rate smoothing $\rho_r$</td>
<td>Beta (0.7,0.1)</td>
</tr>
<tr>
<td>Response to inflation $\rho_{\pi}$</td>
<td>IGamma (1.5,0.1)</td>
</tr>
<tr>
<td>Response to output growth $\rho_y$</td>
<td>IGamma (0.5,0.1)</td>
</tr>
</tbody>
</table>

### Table 3: Priors of shock parameters

<table>
<thead>
<tr>
<th>Priors</th>
<th>Priors</th>
</tr>
</thead>
<tbody>
<tr>
<td>Technology growth $\rho_z$</td>
<td>Beta (0.3,0.1)</td>
</tr>
<tr>
<td>$100\sigma_z$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Monetary policy $\rho_{mp}$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>$100\sigma_{mp}$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Investment $\rho_{\mu}$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>$100\sigma_{\mu}$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Risk premium $\rho_b$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>$100\sigma_b$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Matching efficiency $\rho_\zeta$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>$100\sigma_\zeta$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Price markup (rescaled) $\rho_{\theta^*}$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>$100\sigma_{\theta^*}$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Bargaining power (rescaled) $\rho_{\eta^*}$</td>
<td>Beta (0.5,0.2)</td>
</tr>
<tr>
<td>$100\sigma_{\eta^*}$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Government spending $\rho_g$</td>
<td>Beta (0.7,0.1)</td>
</tr>
<tr>
<td>$100\sigma_g$</td>
<td>IGamma (0.1,3)</td>
</tr>
<tr>
<td>Table 4: Parameters derived from steady-state conditions</td>
<td></td>
</tr>
<tr>
<td>---------------------------------</td>
<td>-----------------</td>
</tr>
<tr>
<td>Employment rate</td>
<td>$N$</td>
</tr>
<tr>
<td>Vacancy</td>
<td>$V$</td>
</tr>
<tr>
<td>Matches</td>
<td>$m$</td>
</tr>
<tr>
<td>Discount factor</td>
<td>$\beta$</td>
</tr>
<tr>
<td>Job survival rate</td>
<td>$\chi$</td>
</tr>
<tr>
<td>Mean of exogenous spending shock</td>
<td>$\epsilon_g$</td>
</tr>
<tr>
<td>Real marginal cost</td>
<td>$\xi$</td>
</tr>
<tr>
<td>Quarterly net real rental rate of capital</td>
<td>$\tilde{r}^K$</td>
</tr>
<tr>
<td>Capital utilization cost first parameter</td>
<td>$\phi_u$</td>
</tr>
<tr>
<td>Capital/output ratio</td>
<td>$k/y$</td>
</tr>
<tr>
<td>Investment/capital ratio</td>
<td>$i/k$</td>
</tr>
<tr>
<td>Investment/output ratio</td>
<td>$i/y$</td>
</tr>
<tr>
<td>Consumption/output ratio</td>
<td>$c/y$</td>
</tr>
<tr>
<td>Pool of job seekers</td>
<td>$S$</td>
</tr>
<tr>
<td>Matching function efficiency</td>
<td>$\zeta$</td>
</tr>
<tr>
<td>Job finding rate</td>
<td>$s$</td>
</tr>
<tr>
<td>Employees' share of output</td>
<td>$\tilde{w}N/y$</td>
</tr>
<tr>
<td>Effective bargaining power</td>
<td>$\eta$</td>
</tr>
<tr>
<td>Autocorrelation of (non-rescaled) markup shock</td>
<td>$\rho_\theta$</td>
</tr>
<tr>
<td>Std dev of (non-rescaled) markup shock</td>
<td>$\sigma_\theta$</td>
</tr>
<tr>
<td>Autocorrelation of (non-rescaled) bargaining power shock</td>
<td>$\rho_\eta$</td>
</tr>
<tr>
<td>Std dev of (non-rescaled) bargaining power shock</td>
<td>$\sigma_\eta$</td>
</tr>
</tbody>
</table>

- $N = 1 - U$  
- $V = \frac{\rho N}{\nu}$  
- $m = qV$  
- $\beta = \frac{\tau}{N}$  
- $\chi = 1 - \rho$  
- $\epsilon_g = \frac{1}{1-g/y}$  
- $\xi = \frac{\theta + 1}{\beta}$  
- $\tilde{r}^K = \frac{\xi}{\beta} - 1 + \delta$  
- $\phi_u = \tilde{r}^K$  
- $\frac{k}{y} = \frac{\alpha \xi}{\tilde{r}^K}$  
- $i = \frac{i}{k}$  
- $i = \frac{i}{k}$  
- $c = \frac{c}{y}$  
- $S = 1 - \chi N$  
- $\zeta = q \left(\frac{V}{N}\right)^\sigma$  
- $s = \zeta \left(\frac{V}{S}\right)^{1-\sigma}$  
- $\tilde{w}N/y = \xi (1 - \alpha) - \frac{(1 - \beta) x_2 \left(\frac{x}{N}\right)^2}{\frac{2}{\tilde{w}N}}$  
- $\eta = \frac{1 - \tau}{\vartheta}$ where $\vartheta \equiv \frac{\xi (1 - \alpha) + (1 + \beta \chi \frac{\xi}{\beta}) x_2 \left(\frac{x}{N}\right)^2}{\frac{2}{\tilde{w}N}}$  
- $\eta = \frac{\eta}{1 - \eta}$  
- $\rho_\theta = \rho_\theta^*$  
- $\sigma_\theta = [(1 + \beta \zeta) \phi_p] \sigma_\theta^*$  
- $\rho_\eta = \rho_\eta^*$  
- $\sigma_\eta = (1 - \eta) \sigma_\eta^*$
3.3 Data transformation

$X_t$ is the vector of observables at time $t$. $X_t$ is expressed in logarithmic deviations from sample mean. $X_t$ contains eight variables: the quarterly growth rate of output, the quarterly growth rate of consumption, the quarterly growth rate of investment, the quarterly growth rate of real wages, the vacancy rate, the unemployment rate, the quarterly inflation rate and the quarterly gross nominal interest rate.

$$X_t = \begin{bmatrix}
\ln (Y_t) - \ln (Y_{t-1}) - \ln(g_y) \\
\ln (C_t) - \ln (C_{t-1}) - \ln(g_c) \\
\ln (I_t) - \ln (I_{t-1}) - \ln(g_i) \\
\ln (W_t) - \ln (W_{t-1}) - \ln(g_w) \\
\ln (V_t) - \ln(V) \\
\ln (U_t) - \ln(U) \\
\ln (P_t) - \ln (P_{t-1}) - \ln(g_p) \\
\ln (R_t) - \ln(R)
\end{bmatrix}.$$  

$Y_t$ is the level of real GDP per capita, $C_t$ is the level of real consumption per capita, $I_t$ is the level of real investment per-capita, $W_t$ is the real wage, $V_t$ is the ratio of the vacancy series constructed by Barnichon (2010) to the sum of the vacancy series and the number of employed people (cf. Justiniano and Michelacci, 2011), $P_t$ is the level of the GDP deflator and $R_t$ is the gross effective federal funds rate, expressed on a quarterly basis. Following the arguments in Shimer (2005), we are detrending the vacancy rate with an HP filter with a smoothing weight equal to $10^{-6}$ to remove the secular trend in the series (cf. also Justiniano and Michelacci 2011 and Davis, Faberman and Haltiwanger 2013).
4 Additional details on the propagation of shocks

In the main text we have concentrated our attention on the transmission mechanism for matching efficiency shocks. In this section we comment on the dynamics induced by the other shocks that are relatively standard. In Figure A0 we plot the responses of the actual and natural rates of unemployment to the six shocks that affect the natural rate. The natural rate of unemployment is defined as the counterfactual rate of unemployment that emerges in the presence of flexible prices and wages and thus corresponds to the concept of unemployment in Real Business Cycle models (Shimer 2005).

The responses of the actual rate are in line with the previous literature. Unemployment is countercyclical in response to all shocks. A partial exception is the case of the neutral technology shocks: on impact (and only on impact) an expansionary technology shock increases unemployment. This is a standard result in New Keynesian models due to the presence of nominal and real rigidities (cf. Galí 1999).

The natural rate does not react to monetary policy and risk premium shocks. It is well known that these shocks propagate only in the presence of nominal rigidities. The natural rate of unemployment reacts little also to technology and investment specific shocks. This result is also well known in the literature since Shimer (2005) and the following literature on the unemployment volatility puzzle. Notice that the nominal rigidities deliver a substantial propagation to these disturbances, thus meaning that the actual rate of unemployment is immune to the unemployment volatility puzzle. In contrast, the natural rate reacts little to technology and investment specific shocks, in line with the measures of natural rates obtained with statistical methods. In the absence of nominal rigidities, an exogenous increase in government spending leads to a very small rise in the unemployment rate. The negative wealth effect triggered by the fiscal impulse generates a fall in consumption and a rise in the real interest rate. Higher real interest rates provide firms with an incentive to raise the rate of capacity utilization, thereby substituting capital services for labor. This channel is amplified by the inelasticity of labor supply in the search and matching model.

As discussed in the main text, the matching efficiency shock has a larger effect on the natural rate than on the actual rate, unlike all the other shocks. This explains why the natural rate is driven almost exclusively by the matching efficiency shock.

In Figure A1 we plot simulated data on vacancies and unemployment conditional on each kind of disturbances. In each panel, the vertical and the horizontal axis correspond respectively to the vacancy rate and the unemployment rate, both expressed in percentage deviations from steady state. Each panel plots pseudo-data points simulated from the model calibrated at the posterior mode and drawing the i.i.d. innovations from normal distributions with mean zero and standard deviation set equal to the corresponding posterior mode estimate. We remark that only the mismatch shock generates a positive conditional correlation between unemployment and vacancies. This point is discussed in detail in the main text and is related to the presence of sticky prices and a pre-match component in total hiring costs. In the data unemployment and vacancies are strongly negatively correlated and, therefore, the other shocks have a better chance to explain aggregate dynamics. Nevertheless, mismatch shocks may play a role in periods when unemployment and vacancies move together.

In Figure A2 we plot the contribution of each shock to the Beveridge curve dynamics. The grey dots represents the dynamics induced by all the eight shocks together. The black dots show how each shock in isolation has moved the Beveridge curve over the period 2008:Q1-2013:Q2. Mismatch shocks have shifted the Beveridge curve to the right. Notice, however, that also other shocks explain part of the shift. All shocks are able to generate the loop typical of Beveridge curve dynamics in recent years and do not generate trajectories along a line. This point has been emphasized by Christiano, Eichenbaum and Trabandt (2014) in a recent paper. However, mismatch shocks are very important to match the shift to the right from a quantitative point of view and more so in recent years. Notice the large effects induced (in opposite directions) by risk
5 Sensitivity analysis

In this section we provide additional details on the sensitivity analysis that we conduct to investigate the robustness of our results. We modify the model along three dimensions: i) the sample period for estimation, ii) the calibration for the elasticity of the matching function to unemployment, iii) the role of a time-varying separation rate. We describe each experiment in turn.

5.1 Sample period

In our baseline model the sample period used for estimation is 1957:Q1-2008:Q4. We now want to investigate the robustness of our results when we consider a longer sample (thus including the Great Recession) and a shorter sample (only the Great Moderation period).

In the first experiment we extend our sample period until 2013:Q2 to exploit the information on the recent shift of the Beveridge curve for estimation purposes. In Figures A3 to A8 we present our results for the extended sample. Matching efficiency is slightly more volatile (Figure A3) than in our baseline estimates but all in all these figures are almost identical to the ones for the baseline case.

In the second extension we focus on a shorter but more homogenous period as the Great Moderation (1984:Q1-2008Q:4). Our baseline sample period is long and may be subject to structural breaks. In contrast, the Great Moderation period is a period of relative stability that may be useful as a cross-check. In Figures A9 to A15 we present the results related to this experiment. Once again all our results on the role of matching efficiency shocks are confirmed. The only difference that we can identify with respect to the baseline case is that the relative importance of the other shocks change slightly, in particular for the risk premium shock. This point can be seen when comparing Figure A2 to Figure A12. However, even from a quantitative point of view these differences are minor. To sum up we conclude that the choice of the sample period for estimation purposes is largely inconsequential.

5.2 Alternative calibration of the matching function elasticity

A key parameter that affects directly the estimated series for matching efficiency shocks is the elasticity of the matching function to unemployment ($\sigma$). In our baseline model we calibrate it at 0.65, a value in the middle of the range (0.55-0.75) found in a series of recent studies (Barnichon and Figura 2014; Justiniano and Michelacci 2011; Shimer 2005; Sedlacek 2014). These values are slightly higher than the ones advocated by Petrongolo and Pissarides (2001) and much higher than the value of 0.4 used by Blanchard and Diamond (1989). Given the importance of this parameter, we reestimate our model with $\sigma$ equal to 0.55 (almost at the bottom of the Petrongolo and Pissarides’ range) and with $\sigma$ equal to 0.75 as in Justiniano and Michelacci (2011).

We plot the estimated series for matching efficiency shocks with $\sigma$ calibrated at 0.55 in Figure A16. In our baseline case (Figure 3 in the main text) matching efficiency increases during some Recessions and declines in others. With $\sigma$ equal to 0.55 matching efficiency becomes more countercyclical: it now often increases during Recessions with the clear exception of the Great Recession when we still identify a substantial decline, followed by a partial rebound and a new and even more pronounced decline. A different series for matching efficiency translates into a different estimate for the natural rate of unemployment given the prominent role of mismatch shocks in its dynamics. In Figure A17 we see that the low frequency dynamics of the

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Footnote: In our model this parameter should be called elasticity of the matching function to searchers since the pool of searchers is not equivalent to unemployment.
natural rate are not affected. However, at high frequencies the correlation between the actual rate and the natural rate is now lower. The natural rate still increases during the Great Recession and keeps increasing in the aftermath as in our baseline case. Mismatch shocks are now less important to explain the shift in the Beveridge curve during the Great Recession (cf. Figure A19 and Figure 5 in the main text) but they are still crucial to explain why unemployment was so high in recent years. Mismatch shocks are still the dominant drivers of the natural rate as it can be seen in Figure A21. We conclude that our main results are confirmed but a low value of $\sigma$ impacts the estimate of matching efficiency and the behavior of natural rate at high frequencies.

Not surprisingly we find the opposite results with a high value of $\sigma$. When $\sigma$ is calibrated at 0.75, matching efficiency declines in almost all Recessions (thus becoming very procyclical, cf. Figure A22) and the natural rate of unemployment becomes more correlated with the actual rate at high frequencies (cf. Figure A23). Matching efficiency shocks are now crucial to explain the Beveridge curve dynamics both during the Great Recession and in its aftermath (cf. Figure A26).

5.3 Time-varying separation rate

Separation shocks instead of mismatch shocks. In this last set of experiments we consider exogenous shocks to the separation rate. Hosios (1994) and Shimer (2005) among others have shown that shocks to the separation rate are also able to move unemployment and vacancies in the same direction. In a first experiment we estimate a model featuring separation shocks rather than mismatch shocks over the sample period 1957:Q1-2013:Q2 to include as many Beveridge curve shifts as possible. We use the same 8 observable variables that we have used in our baseline model. In Figure A28 we plot the estimated series for the separation shock. It is almost the mirror image of the matching efficiency series, reflecting the fact that two shocks are nearly observationally equivalent. The estimate for the natural rate of unemployment does not change significantly with two exceptions: it is barely above the actual rate in the pre-Great Recession period and it is above the actual rate in the last quarters of the sample. Separation shocks are now the dominant drivers of the natural rate (Figure A33) and are essential to explain the shift of the Beveridge curve in recent years (Figure A32). Unlike mismatch shocks, separation shocks propagate more in the model with post-match hiring costs only (Figure A35). However, at the posterior mode value for the share of pre-match hiring costs ($\phi_V$) the propagation of the two shocks is similar. Separation shocks also propagate more in the model with flexible prices and wages and the difference with the benchmark with nominal rigidities is even larger than in the case of mismatch shocks (Figure A36).

We conclude that mismatch and separation shocks play a very similar role in our estimated model and that both shocks may explain a shift in the Beveridge curve. However, while several studies find evidence for a decline in matching efficiency in recent years, the estimated separation shock is at odd with JOLTS data in which, somewhat surprisingly, the separation rate is constant throughout the Great Recession and declines afterwards. Layoffs increased sharply during the crisis but quits declined by a comparable amount leaving the aggregate separation rate almost unaffected. In more recent years layoffs declined to pre-crisis levels while the increase in quits has been moderate, thus leaving the separation rate still well below its pre-crisis value (cf. Christiano, Eichenbaum and Trabandt, 2014). The discrepancy between the JOLTS data and the estimated separation shock suggests that, at least during the Great Recession, other factors than separation were at work and contributed to the shift in the Beveridge curve.

Separation rate correlated with the state of the economy. The results from the previous exercise suggest that mismatch shocks and separation shocks are nearly observationally equivalent. To overcome this issue, we first relax the assumption of constant separation without including exogenous shocks to the separation rate. We assume that the separation rate is negatively related to the state economy (i.e. the
separation rate is low in good times) where the state of the economy is summarized by the technology and the investment-specific shocks, the two main drivers of business cycle fluctuations in our model. We assume the following specification:

\[
\ln \rho_t = (1 - \rho_{\rho}) \ln \rho + \rho_{\rho} \ln \rho_{t-1} - \delta_{z} \varepsilon_{zt} - \delta_{\mu} \varepsilon_{\mu t}
\]

where we impose in the estimation that \( \delta_{z} \) and \( \delta_{\mu} \) have to be positive. With this specification of the separation rate, it is challenging to identify separately the matching shock parameters and the new parameters governing the evolution of the separation rate. Hence, to improve the identification of these parameters, we fix (calibrate) the deep parameters at a value roughly in line with the estimates obtained for our baseline model over the longest sample period. This implies that we only estimate the parameters of the shock processes. The priors on the new parameters \( \delta_{z} \) and \( \delta_{\mu} \) are Uniform.

In Figures A38 to A43 we present graphically our results. We still identify a large decline in matching efficiency during the Great Recession and a zero unemployment gap at the end of the sample. Moreover, matching efficiency shocks are still important to explain the recent shift in the Beveridge curve. A series of negative matching efficiency shocks contribute to slowdown the recovery in the period 2011-2013. Technology shocks and investment-specific shocks play now a somewhat larger role in the dynamics of the natural rate of unemployment but the mismatch shock is still the most important driver.

Mismatch shocks and separation shocks together. In our last robustness check we consider separation shocks and matching efficiency shocks in the same estimation exercise, unlike in the previous experiments. To achieve identification we need an additional observable variable which is the separation rate, constructed by converting monthly JOLTS data on total separation in the non-farm private sector to quarterly frequency. The JOLTS separation rate exhibits a clear downward trend over the whole period for which this series is available, namely since December 2000 until the end of the sample. Hence, we detrend the separation rate by applying the HP filter with a smoothing weight of \( 10^{-6} \). We still allow the separation rate to be contemporaneously negatively correlated with the neutral technology and the investment-specific innovations, as it would be in a model with endogenous separation. The separation rate follows now the following process:

\[
\ln \rho_t = (1 - \rho_{\rho}) \ln \rho + \rho_{\rho} \ln \rho_{t-1} - \delta_{z} \varepsilon_{zt} - \delta_{\mu} \varepsilon_{\mu t} + \varepsilon_{\rho t}
\]

where the new element compared to the previous case is given by the presence of the separation shock \( \varepsilon_{\rho t} \). The sample period of our dataset with nine observables is considerably shorter than in all our previous robustness checks (JOLTS data are available only since 2000). Therefore, we calibrate all the deep parameters at a value in line with our baseline model estimated over the period 1957:Q1-2013:Q2. Once again, here we only estimate the parameters of the shock processes.

We find that our results are confirmed also in this case where mismatch shocks and separation coexist. The results for this version of the model with nine observables and nine shocks are presented in Figures A44 to A49. We find once again evidence of a material deterioration in matching efficiency during the Great Recession, with matching efficiency recovering only extremely slowly from 2010 to mid 2013. This decline in matching efficiency has a long-lasting impact on unemployment throughout the recovery, contributing to lift the unemployment rate by about one percent on average over the period 2010-2013, a slightly lower effect than in our baseline model. The impact on the natural rate is confirmed: The decline in matching efficiency generates an increase in the natural rate of unemployment of about 1 percentage point on average over the
period beginning of 2010-mid 2013, again somewhat lower than in the baseline model. Separation shocks are now also important drivers of the natural rate but contribute to lower it over the period 2009-2012. This implies that mismatch shocks are still very important to explain the shift to the right of the Beveridge curve. While expansionary separation shocks were shifting the curve to the left (cf. Figure A49), large negative matching efficiency shocks have produced a shift in the opposite direction, thus allowing the model to match the data.

5.4 Summary

We conclude that when we change the sample period, the calibration for the elasticity of the matching function to unemployment or when we consider shocks to the separation rate, all our main results are confirmed. Matching efficiency shocks are not important drivers of the business cycle but they may play a role in selected periods and they are the most important driver of the natural rate (even in the presence of separation shocks). According to our analysis, they contribute substantially to explain the shift of the Beveridge curve and the weak recovery in the aftermath of the Great Recession. A model with separation shocks (and without matching efficiency shocks) can deliver similar results but implies a series for the separation shock that is in strong contrast with the JOLTS data.
References


Fig. A0: Impulse responses of the actual and natural unemployment rates, expressed in percentage points. The responses are computed at the posterior mode. The size of each shock is one standard deviation.
Fig A1: Simulated conditional Beveridge curves
Risk premium shocks

Matching shocks

Technology shocks

Investment shocks

Markup shocks

Bargaining shocks

Monetary shocks

Fiscal shocks

Fig A2: Contribution of each shock to the Beveridge curve 2008Q1-2013Q2 (% dev. from 2008Q1).
Robustness Check #1 - Estimation Period: 1957:Q1 - 2013:Q2

Calibrated Parameters: Check #1 [57:Q1-13:Q2]

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
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<tbody>
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<tr>
<td>Capital share</td>
<td>( \alpha )</td>
</tr>
<tr>
<td>Elasticity of substitution btw goods</td>
<td>( \theta )</td>
</tr>
<tr>
<td>Backward-looking price setting</td>
<td>( \varsigma )</td>
</tr>
<tr>
<td>Replacement rate</td>
<td>( \tau )</td>
</tr>
<tr>
<td>Job destruction rate</td>
<td>( \rho )</td>
</tr>
<tr>
<td>Elasticity of matches to unemp.</td>
<td>( \sigma )</td>
</tr>
<tr>
<td>Probability to fill a vacancy within a quarter</td>
<td>( q )</td>
</tr>
<tr>
<td>Exogenous spending/output ratio</td>
<td>( g/y )</td>
</tr>
<tr>
<td>Unemployment rate</td>
<td>( U )</td>
</tr>
<tr>
<td>Quarterly gross growth rate</td>
<td>( z )</td>
</tr>
<tr>
<td>Quarterly gross inflation rate</td>
<td>( \pi )</td>
</tr>
<tr>
<td>Quarterly gross nominal interest rate</td>
<td>( R )</td>
</tr>
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Priors and Posteriors: Check #1 [57:Q1-13:Q2]

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<th>Prior Distribution</th>
<th>Post. Mode</th>
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<tr>
<td>Weight of pre-match cost in total hiring cost</td>
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<td>( h ) Beta (0.7,0.1)</td>
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<td>( \phi_P ) IGamma (60,10)</td>
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<td>( \phi_W ) IGamma (150,25)</td>
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<tr>
<td>Wage indexation</td>
<td>( \varrho ) Beta (0.5,0.2)</td>
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<tr>
<td>Interest smoothing</td>
<td>( \rho_r ) Beta (0.7,0.1)</td>
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</tr>
<tr>
<td>Resp. to inflation</td>
<td>( \rho_\pi ) IGamma (1.5,0.1)</td>
<td>1.97</td>
</tr>
<tr>
<td>Resp. to growth</td>
<td>( \rho_y ) IGamma (0.5,0.1)</td>
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Robustness Check #2 - Estimation Period: 1985:Q1 - 2008:Q3

Calibrated Parameters: Check #2 [85:Q1-08:Q3]

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<tr>
<td>Capital share</td>
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<tr>
<td>Elasticity of substitution btw goods</td>
<td>θ 6</td>
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<tr>
<td>Backward-looking price setting</td>
<td>ς 0.01</td>
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<tr>
<td>Replacement rate</td>
<td>τ 0.40</td>
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<tr>
<td>Job destruction rate</td>
<td>ρ 0.085</td>
</tr>
<tr>
<td>Elasticity of matches to unemp.</td>
<td>σ 0.65</td>
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<tr>
<td>Probability to fill a vacancy within a quarter</td>
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<tr>
<td>Exogenous spending/output ratio</td>
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<td>Unemployment rate</td>
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<tr>
<td>Quarterly gross growth rate</td>
<td>z 1.0043</td>
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<tr>
<td>Quarterly gross inflation rate</td>
<td>π 1.0061</td>
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<tr>
<td>Quarterly gross nominal interest rate</td>
<td>R 1.0125</td>
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Priors and Posteriors: Check #2 [85:Q1-08:Q3]

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<td>φ_u2 IGamma (0.5,0.1)</td>
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<td>Price adjust. cost</td>
<td>φ_P IGamma (60,10)</td>
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<td>φ_W IGamma (150,25)</td>
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<td>ρ_r Beta (0.7,0.1)</td>
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<td>Resp. to inflation</td>
<td>ρ_π IGamma (1.5,0.1)</td>
<td>2.33</td>
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<tr>
<td>Resp. to growth</td>
<td>ρ_y IGamma (0.5,0.1)</td>
<td>0.41</td>
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Robustness Check #3 - $\sigma = 0.55$ (Est. Per.: 57:Q1 - 08:Q3)

### Calibrated Parameters: Check #3 Low Sigma

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<td>Elasticity of substitution btw goods $\theta$</td>
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<td>Backward-looking price setting $\varsigma$</td>
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<tr>
<td>Replacement rate $\tau$</td>
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<tr>
<td>Job destruction rate $\rho$</td>
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<tr>
<td>Quarterly gross inflation rate $\pi$</td>
<td>1.0088</td>
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<tr>
<td>Quarterly gross nominal interest rate $R$</td>
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### Priors and Posteriors: Check #3 Low Sigma

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<th>Post. Mode</th>
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<tr>
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<td>Resp. to growth $\rho_y$</td>
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Robustness Check #4 - $\sigma = 0.75$ (Est. Per.: 57:Q1 - 08:Q3)

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<td>Capital share</td>
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<tr>
<td>Elasticity of substitution btw goods</td>
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<tr>
<td>Backward-looking price setting</td>
</tr>
<tr>
<td>Replacement rate</td>
</tr>
<tr>
<td>Job destruction rate</td>
</tr>
<tr>
<td>Elasticity of matches to unemp.</td>
</tr>
<tr>
<td>Probability to fill a vacancy within a quarter</td>
</tr>
<tr>
<td>Exogenous spending/output ratio</td>
</tr>
<tr>
<td>Unemployment rate</td>
</tr>
<tr>
<td>Quarterly gross growth rate</td>
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<tr>
<td>Quarterly gross inflation rate</td>
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<tr>
<td>Quarterly gross nominal interest rate</td>
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<table>
<thead>
<tr>
<th>Priors and Posteriors: Check #4 - High Sigma</th>
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<tbody>
<tr>
<td>Weight of pre-match cost in total hiring cost</td>
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<td>Hiring cost/output ratio</td>
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<td>Habit in consump.</td>
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<tr>
<td>Invest. adj. cost</td>
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<tr>
<td>Capital ut. cost</td>
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<tr>
<td>Price adjust. cost</td>
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<tr>
<td>Wage adjust. cost</td>
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<tr>
<td>Wage indexation</td>
</tr>
<tr>
<td>Interest smoothing</td>
</tr>
<tr>
<td>Resp. to inflation</td>
</tr>
<tr>
<td>Resp. to growth</td>
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</table>
Fig A22: Matching Efficiency (\( \sigma = 0.75 \); estim period: 57-08Q3)

Fig A23: Unemployment Rates: Actual vs Natural (\( \sigma = 0.75 \); estim period: 57-08Q3)

Fig A24: Unemployment Gap: Median and 90% Bands (\( \sigma = 0.75 \); estim period: 57-08Q3)

Calibrated Parameters: Check #5 - Only separ. shock

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<tr>
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<td>α 0.33</td>
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<tr>
<td>Elasticity of substitution btw goods</td>
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<tr>
<td>Backward-looking price setting</td>
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<td>Replacement rate</td>
<td>τ 0.40</td>
</tr>
<tr>
<td>Job destruction rate</td>
<td>ρ 0.085</td>
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<tr>
<td>Elasticity of matches to unemp.</td>
<td>σ 0.65</td>
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<tr>
<td>Probability to fill a vacancy within a quarter</td>
<td>q 0.70</td>
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<tr>
<td>Exogenous spending/output ratio</td>
<td>g/y 0.20</td>
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Priors and Posteriors: Check #5 - Only separ. shock

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<td>Habit in consump.</td>
<td>h Beta (0.7,0.1)</td>
<td>0.64</td>
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<td>ϕ_I IGamma (5,1)</td>
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<td>ϕ_u2 IGamma (0.5,0.1)</td>
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<td>ϕ_P IGamma (60,10)</td>
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<tr>
<td>Interest smoothing</td>
<td>ρ_r Beta (0.7,0.1)</td>
<td>0.36</td>
</tr>
<tr>
<td>Resp. to inflation</td>
<td>ρ_π IGamma (1.5,0.1)</td>
<td>1.91</td>
</tr>
<tr>
<td>Resp. to growth</td>
<td>ρ_y IGamma (0.5,0.1)</td>
<td>0.37</td>
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Fig. A28 - A33: Model featuring a separation shock instead of the matching efficiency shock. This version of the model is estimated using the same 8 observables as in our baseline, over the period 1957Q1 - 2013Q2.
Fig. A31 - A33: Model featuring a separation shock instead of the matching efficiency shock. Model estimated using the same 8 observables as in our baseline case, over the period 1957Q1-2013Q2.
Fig. A34: Conditional Beveridge curves. Model featuring a separation shock instead of the matching shock. Estimation period: 1957Q1 - 2013Q2, using the same 8 observables as in our baseline case.
Fig. A35: Impulse responses to a one-standard-deviation separation rate shock. Model features a separation shock instead of the matching efficiency shock. Impulse responses computed for different values of the pre-match hiring cost weight $\phi_V$ and keeping all other parameters at the posterior mode. Model estimated using standard 8 observables over period 1957Q1-2013Q2.
Fig. A36: Impulse responses to a one-standard-deviation separation shock in the actual model economy with nominal rigidities and in the counterfactual economy without nominal rigidities. The model features a separation shock instead of the matching shock and is estimated using the same 8 observables as in our benchmark case, over period 1957Q1 - 2013Q2. Impulse responses computed at the posterior mode.
Fig. A37: Simulated Beveridge curves conditional on one kind of shock at a time. The model features a separation shock instead of the matching shock and has been estimated using the usual 8 observables over the period 1957Q1 - 2013Q2. All parameters set at posterior mode.
#6: Model with separation rate correlated to tech and IST shocks

- Sep. rate follows: $\ln r_t = (1 - \rho_\rho) \ln r + \rho_\rho \ln r_{t-1} - \delta z_t - \delta \mu_\mu$, where $\delta_z \geq 0$, $\delta_\mu \geq 0$.
- Only shocks’ parameters are estimated. Use Uniform priors for $\delta_z \in (0, 3)$ and $\delta_\mu \in (0, 1)$.

### Calibrated Parameters: Check #6 - Time-varying separation rate

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<td>Capital share</td>
<td>$\alpha$ 0.33</td>
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<td>Elasticity of substitution btw goods</td>
<td>$\theta$ 6</td>
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<tr>
<td>Backward-looking price setting</td>
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<tr>
<td>Replacement rate</td>
<td>$\tau$ 0.40</td>
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<tr>
<td>Job destruction rate</td>
<td>$\rho$ 0.085</td>
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<td>Elasticity of matches to unemp.</td>
<td>$\sigma$ 0.65</td>
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<td>$q$ 0.70</td>
</tr>
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<td>Exogenous spending/output ratio</td>
<td>$g/y$ 0.20</td>
</tr>
<tr>
<td>Unemployment rate</td>
<td>$U$ 0.0602</td>
</tr>
<tr>
<td>Quarterly gross growth rate</td>
<td>$z$ 1.0038</td>
</tr>
<tr>
<td>Quarterly gross inflation rate</td>
<td>$\pi$ 1.0083</td>
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<tr>
<td>Quarterly gross nominal interest rate</td>
<td>$R$ 1.0128</td>
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<tr>
<td>Weight of pre-match cost in total hiring cost</td>
<td>$\phi_v$ 0.25</td>
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<tr>
<td>Hiring cost/output ratio</td>
<td>$1000 \frac{\phi_v}{\phi_{\mu^2}}$ 2.50</td>
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<tr>
<td>Habit in consmp.</td>
<td>$h$ 0.64</td>
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<tr>
<td>Invest. adj. cost</td>
<td>$\phi_I$ 3.00</td>
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<tr>
<td>Capital ut. cost</td>
<td>$\phi_{\mu^2}$ 0.60</td>
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<tr>
<td>Price adjust. cost</td>
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<tr>
<td>Wage adjust. cost</td>
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<td>Wage indexation</td>
<td>$\rho$ 0.96</td>
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<tr>
<td>Interest smoothing</td>
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<td>Resp. to inflation</td>
<td>$\rho_\pi$ 1.90</td>
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Fig. A38 - A43: Model where separation rate is contemporaneously correlated with neutral technology and IST innovations. Only the shocks parameters are estimated using the usual 8 observables over 57Q1-13Q2.
Fig A40: Beveridge Curve (estim. period: 57Q1-13Q2)

Fig A41: Historical Decomp. of Unemp. Rate (estim. period: 57Q1-13Q2)

Fig A42: Historical Decomp. of Natural Rate (estim. period: 57Q1-13Q2)

Fig A38 - A43: Model where separation rate is contemporaneously correlated with neutral technology and IST innovations. Only the shocks parameters are estimated using the usual 8 observables over 57Q1-13Q2.
Fig. A43: Conditional Beveridge curves. Model with time-varying separation rate correlated with contemporaneous techno and IST innovations. Only shocks param were estimated using usual 8 observables over 1957:Q1 - 2013:Q2.
#7: Model with matching and separation shocks
- 9 Observables including JOLTS separation rate (Total Sep. Non Farm) over 01:Q1-13:Q2.
- Convert monthly sep. rate to quarterly and detrend with HP filter ($\lambda = 10^6$).
- Sep. rate follows: $\ln p_t = (1 - \rho) \ln p_{t-1} - \delta \varepsilon_{zt} - \delta \mu \varepsilon_{\mu t} + \varepsilon_{pt}, \delta_z \geq 0, \delta_{\mu} \geq 0$.
- Only shocks' parameters are estimated. Use Uniform priors for $\delta_z \in (0,3)$ and $\delta_{\mu} \in (0,1)$.

Calibrated Parameters: Check #7 - Both match & separ shocks

<table>
<thead>
<tr>
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<td>Replacement rate</td>
<td>$\tau$ 0.40</td>
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Fig. A44: Matching Efficiency and Separation Rate

Fig. A44 - A49: We first convert JOLTS monthly data on "Total separation rate: Private - Non farm" to quarterly frequency. We then detrend the quarterly rate using the HP filter with a smoothing weight equal to 1,000,000.

Fig. A45: Unemployment Rates: Actual vs Natural

Fig. A44 - A49: Model featuring both a matching shock and a separation shock, estimated using 9 observables (usual 8 + separation rate), over 2001:Q1 - 2013:Q2. Only shocks' parameters are estimated.
Fig A46: Beveridge Curve (estim. period: 01Q1-13Q2)

Fig A47: Historical Decomp. of Unemp. Rate (estim. period: 01Q1-13Q2)

Fig A48: Historical Decomp. of Natural Rate (estim. period: 01Q1-13Q2)

Fig. A44 – A49: Model with both matching and separation shocks, estimated using 9 observables (usual 8 + sep. rate) over 2000:Q1 – 2013:Q2. Only shocks’ parameters are estimated.
Fig. A49: Conditional Beveridge curves. Model with both matching and separation shocks estimated with 9 observables over 2001:Q1 - 2013:Q2. Only shocks’ parameters were estimated.