The optimal Inflation Target and the Natural Rate of Interest ¹

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January 10, 2018

Abstract

We study how changes in the value of the steady-state real interest rate affect the optimal inflation target, both in the U.S. and the euro area, using an estimated New Keynesian DSGE model that incorporates the zero (or effective) lower bound on the nominal interest rate. We find that this relation is downward sloping, but its slope is not necessarily one-for-one: increases in the optimal inflation rate are generally lower than declines in the steady-state real interest rate. Our approach allows us not only to assess the uncertainty surrounding the optimal inflation target, but also to determine the latter while taking into account the parameter uncertainty facing the policy maker, including uncertainty with regard to the determinants of the steady-state real interest rate. We find that in the currently empirically relevant region for the US as well as the euro area, the slope of the curve is close to -0.9. That finding is robust to allowing for parameter uncertainty.

Keywords: inflation target, Effective lower bound.
JEL Codes: E31, E52, E58

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1 Introduction

A recent but sizable literature has pointed to a permanent—or, at least very persistent—decline in the “natural” rate of interest in advanced economies (Holston et al., 2017, Laubach and Williams, 2016). Various likely sources of that decline have been discussed, including a lower trend growth rate of productivity (Gordon, 2015), or an enhanced preference for safe assets (Caballero and Farhi, 2015, Summers, 2014).

A lower steady-state real interest rate matters for monetary policy. Given average inflation, a lower steady-state real rate will cause the nominal interest rate to hit its zero lower bound (ZLB) more frequently, hampering the ability of monetary policy to stabilize the economy, bringing about more frequent (and potentially protracted) episodes of recessions and below-target inflation.

In the face of that risk, and in order to counteract it, several prominent economists have forcefully argued in favor of raising the inflation target (see, among others, Ball 2014, Blanchard et al. 2010, Williams 2016). Since a lower natural rate of interest is conducive to a higher ZLB incidence, one would expect a higher inflation target to be desirable in that case. But the answer to the practical question of by how much should the target be increased is not obvious. Indeed, the benefit of providing a better hedge against hitting the ZLB, which is an infrequent event, comes at a cost of higher steady-state inflation which induces permanent costs, as recently argued by Bernanke (2016) among others. The answer to this question thus requires to assess how the tradeoff between the incidence of the ZLB and the welfare cost induced by steady-state inflation is modified when the natural rate of interest decreases. While the decrease in the natural rate of interest has been emphasized in the recent literature, this assessment has received surprisingly little attention.

The present paper contributes to this debate by asking two questions. First, to what extent does a lower steady-state real interest rate ($r^*$) call for a higher optimal inflation target ($\pi^*$)? Second, does the source of decline in $r^*$ matter? Our main contribution is to characterize how the optimal inflation target is related to the steady-state real interest rate, using a structural, empirically estimated, macroeconomic model. Our main findings can be summarized as follows: (i) The relation between $r^*$ and $\pi^*$ is downward sloping, but not necessarily one-for-one; (ii) in the vicinity of the pre-crisis values for $r^*$, the slope of the $(r^*, \pi^*)$ locus is close to $-0.9$; and (iii) for a plausible range of $r^*$ values the relation is largely robust to the underlying source of variation in $r^*$.

Our results are obtained from extensive simulations of a New Keynesian DSGE model estimated for both the US and the euro area over a Great Moderation sample. The framework features: (i) price stickiness and imperfect indexation of prices to non-zero trend inflation, (ii) wage stickiness and imperfect indexation of wages to both inflation and technical progress, and (iii) a ZLB constraint on the nominal interest rate. The first two features imply the presence of potentially substantial costs associated with non-zero wage or price inflation. The third feature warrants a strictly positive inflation rate, in order to mitigate the incidence and
adverse effects of the ZLB. To our knowledge, these three features have not been jointly taken into account in previous analyses of optimal inflation.

According to our simulations, the pre-crisis optimal inflation target obtained when the policymaker is assumed to know the economy’s parameters with certainty (and taken to correspond to the mode of the posterior distribution) is around 2% for the US and around 1.5% for the euro area (in annual terms). This result is obtained in an environment with a relatively low probability of hitting the ZLB (6% for the US and slightly less than 10% for the euro area), given the small shocks estimated on our Great Moderation sample.

A further noticeable feature of our approach is that we perform a full-blown Bayesian estimation of the model, using both US and euro area data. This allows us not only to assess the uncertainty surrounding \( \pi^* \), but also to derive an optimal inflation target taking into account the parameter uncertainty facing the policy maker, including uncertainty with regard to the determinants of the steady-state real interest rate. When that parameter uncertainty is allowed for, those values increase significantly: 2.40% for the US and 2.20% for the euro area. The reason why the optimal targets under parameter uncertainty are higher has to do with the fact that the loss function is asymmetric so that choosing an inflation target that is lower than the optimal one is more costly than choosing an inflation target that is above. That being said it remains true that a Bayesian-theoretic optimal inflation target rises by about 90 basis points in response to a downward shift of the distribution in \( r^* \) of 100 basis points.

1.1 Related Literature

To our knowledge no paper has systematically investigated the \((r^*, \pi^*)\) relation. Coibion et al. (2012) (and its follow-up Dordal-i-Carreras et al. (2016)) and Kiley and Roberts (2017) are the papers most closely related to ours, as they study optimal inflation in quantitative set-ups that account for the ZLB. However, their analyses hold for a constant steady-state natural rate of interest. Relative to both papers, a key difference is our focus on eliciting the relation between the steady-state real interest rate and optimal inflation. Other differences are (i) our interest in the euro area, in addition to the US; (ii) we estimate, rather than calibrate, the model, and (iii) we allow for wage rigidity in the form of infrequent, staggered, wage adjustments. A distinctive feature with respect to Kiley and Roberts (2017) is that we use a model-consistent, micro-founded loss function to compute optimal inflation.

A series of papers assessed the probability of hitting the ZLB for a given inflation target (see, among others, Chung et al. 2012, Coenen 2003, Coenen et al. 2004). Interestingly, our own assessment of this pre-crisis ZLB incidence falls broadly in the ballpark of available estimates. A related recent contribution by Gust et al. (2017) emphasizes that the ZLB was indeed a significant constraint on monetary policy that exacerbated the recession and inhibited the recovery.\(^1\)

\(^1\)Gust et al. (2017) rely on global solution methods while we resort to the lighter algorithms developed by Bodenstein et al.
Other relevant references, albeit ones that put little or no emphasis on the ZLB, are the following. Ascari et al. (2015) study optimal inflation in a model with nominal rigidities. As we do, they include wage rigidity, but do not incorporate the ZLB. Khan et al. (2003) was one of the first papers to study optimal inflation with sticky prices, along with demand for money motives. Schmitt-Grohé and Uribe (2010) review various determinants of the long-run inflation rate, including some not considered in the present paper (e.g., transactions frictions, changes in product quality). Amano et al. (2009) show that optimal inflation might be negative in a model with prices and wages both sticky. Adding search and matching frictions to the setup, Carlsson and Westermark (2016) show, by contrast, that optimal inflation can be positive. Bilbiie et al. (2014) find positive optimal inflation can be an outcome in a sticky-price model with endogenous entry and product variety. Somewhat related, Adam and Weber (2017) show that, even without any ZLB concern, optimal inflation might be positive in the context of a model with heterogeneous firms and systematic firm-level productivity trends. Finally, using a perpetual youth model, Lepetit (2017) shows that optimal inflation can be positive in the presence of heterogeneous discount factor, especially when the social planner is more patient than agents.

Among the recent papers with ZLB, Blanco (2016) studies optimal inflation in a state-dependent pricing model, i.e. a “menu cost” model (see also Burstein and Hellwig 2008 for a similar exercise without ZLB, which leads to negative optimal inflation rate). As a matter of fact, Nakamura et al. (2016) argue that the presence of state-dependent pricing limits considerably the positive relationship between inflation and price dispersion, thus limiting the costs of inflation.

2 The Model

We use a relatively standard medium-scale New Keynesian model as a framework of reference. Crucially, the model features elements that generate a cost to inflation: (1) nominal rigidities, in the form of staggered price and wage setting; (2) less than perfect price (and wage) indexation to past or trend inflation; and (3) trend productivity growth along, to which wages are imperfectly indexed.

As is well known, staggered price setting generates a positive relation between deviations from zero inflation and price dispersion (with the resulting inefficient allocation of resources). Moreover, the lack of indexation to trend magnifies these costs, as emphasized by Ascari and Sbordone (2014). Also, and ceteris paribus, price inflation induces (nominal) wage inflation, which in turn triggers inefficient wage dispersion in the presence of staggered wage setting. Imperfect indexation also magnifies the costs of non-zero price (or wage) inflation as compared to a set-up where price and wages mechanically catch up with trend inflation. Finally the lack of a systematic indexation of wages to productivity also induces an
inefficient wage dispersion.

At the same time, there are benefits associated to a positive inflation rate, as interest rates are subject to a ZLB constraint. In particular, and given the steady-state real interest rate, the incidence of binding ZLB episodes should decline with the average rate of inflation. Such episodes hamper the stabilization potential of monetary policy.

Overall, the model we use, and the trade-off between costs and benefits of steady-state inflation, are close to those considered by Coibion et al. (2012). However we assume sticky wages, in addition to sticky prices.

2.1 Households

The economy is inhabited by a continuum of measure one of infinitely-lived, identical households. The representative household is composed of a continuum of workers, each specialized in a particular labor type indexed by $h \in [0,1]$. The representative household’s objective is to maximize an intertemporal welfare function

$$\mathbb{E}_t \sum_{s=0}^{\infty} \beta^s \left\{ e^{\zeta_{g,t+s}} \log(C_{t+s} - \eta C_{t+s-1}) - \frac{X}{1+\nu} \int_0^1 N_{t+s}(h)^{1+\nu} dh \right\},$$

where $\beta \equiv e^{-\rho}$ is the discount factor ($\rho$ being the discount rate), $\mathbb{E}_t \{ \cdot \}$ is the expectation operator conditional on information available at time $t$, $C_t$ is consumption and $N_t(h)$ is the supply of labor of type $h$. The utility function features habit formation, with degree of habits $\eta$. The inverse Frisch elasticity of labor supply is $\nu$ and $\chi$ is a scale parameter in the labor disutility. The utility derived from consumption is subject to a preference shock $\zeta_{g,t}$.

The representative household maximizes (1) subject to the sequence of constraints

$$P_tC_t + e^{\zeta_{q,t}} Q_t B_t \leq \int_0^1 W_t(h) N_t(h) dh + B_{t-1} - T_t + D_t$$

where $P_t$ is the aggregate price level, $W_t(h)$ is the nominal wage rate associated with labor of type $h$, $e^{\zeta_{q,t}} Q_t$ is the price at $t$ of a one-period nominal bond paying one unit of currency in the next period, where $\zeta_{q,t}$ is a “risk-premium” shock, $B_t$ is the quantity of such bonds acquired at $t$, $T_t$ denotes lump-sum taxes, and $D_t$ stands for the dividends rebated to the households by monopolistic firms.

2.2 Firms and Price Setting

The final good is produced by perfectly competitive firms according to the Dixit-Stiglitz production function

$$Y_t = \left( \int_0^1 Y_t\left(f^{(\theta_p-1)/\theta_p} df\right) \right)^{\theta_p/(\theta_p-1)},$$
where \( Y_t \) is the quantity of final good produced at \( t \), \( Y_t(f) \) is the input of intermediate good \( f \), and \( \theta_p \) the elasticity of substitution between any two intermediate goods. The zero-profit condition yields the relation

\[
P_t = \left( \int_0^1 P_t(f)^{1-\theta_p} \, df \right)^{1/(1-\theta_p)}.
\]

Intermediate goods are produced by monopolistic firms, each specialized in a particular good \( f \in [0, 1] \). Firm \( f \) has technology

\[
Y_t(f) = Z_t L_t(f)^{1/\phi}.
\]

where \( L_t(f) \) is the input of aggregate labor, \( 1/\phi \) is the elasticity of production with respect to aggregate labor, and \( Z_t \) is an index of aggregate productivity. The latter evolves according to

\[
Z_t = Z_{t-1} e^{\mu_z + \xi_t},
\]

where \( \mu_z \) is the average growth rate of productivity. Thus, technology is characterized by a unit root in the model.

Intermediate goods producers are subject to nominal rigidities à la Calvo. Formally, firms face a constant probability \( \alpha_p \) of not being able to re-optimize prices. In the event that firm \( f \) is not drawn to re-optimize at \( t \), it re-scales its price according to the indexation rule

\[
P_t(f) = \left[ (\Pi)^{1-\iota_p} (\Pi_{t-1})^{\iota_p} \right]^{\gamma_p} P_{t-1}(f)
\]

where \( \Pi_t \equiv P_t/P_{t-1} \), \( \Pi \) is the associated steady-state value, \( \iota_p \in [0, 1] \) and \( 0 \leq \gamma_p < 1 \). Thus, in case firm \( f \) is not drawn to re-optimize, it mechanically re-scales its price using a geometric average of steady-state inflation and past inflation. Importantly, however, we assume that the degree of indexation is less than perfect since \( \gamma_p < 1 \).

If drawn to re-optimize in period \( t \), a firms chooses \( P_t^* \) in order to maximize

\[
\mathbb{E}_t \sum_{s=0}^{\infty} (\beta \alpha_p)^s \Lambda_{t+s} \left\{ (1 + \tau_{p,t+s}) \frac{V_{t,t+s}^p P_t^*}{P_{t+s}} Y_{t,t+s} - \frac{W_{t+s}}{P_{t+s}} \left( \frac{Y_{t,t+s}}{Z_{t+s}} \right)^\phi \right\},
\]

where \( \Lambda_t \) denotes the marginal utility of wealth, \( \tau_{p,t} \) is a sales tax paid by firms and rebated to households in a lump-sum fashion, and \( Y_{t,t+s} \) is the demand function that a monopolist who last revised its price at \( t \) faces at \( t + s \); it obeys

\[
Y_{t,t+s} = \left( \frac{V_{t,t+s}^p P_t^*}{P_{t+s}} \right)^{\theta_p} Y_{t+s}
\]

where \( V_{t,t+s}^p \) reflects the compounded effects of price indexation to past inflation

\[
V_{t,t+s}^p = \prod_{j=t}^{t+s-1} [(\Pi)^{1-\iota_p} (\Pi_j)^{\iota_p}]^{\gamma_p}.
\]
We further assume that
\[ 1 + \tau_{p,t} = (1 + \tau_p)e^{-\zeta_{u,t}}, \]
with \( \zeta_{u,t} \) appearing in the system as a cost-push shock. Furthermore, we set \( \tau_p \) so as to neutralize the steady-state distortion induced by price markups.

### 2.3 Aggregate Labor and Wage Setting

There is a continuum of perfectly competitive labor aggregating firms that mix the specialized labor types according to the CES technology
\[ N_t = \left( \int_0^1 N_t(h)^{(\theta_w - 1)/\theta_w} \, dh \right)^{\theta_w/(\theta_w - 1)}, \]
where \( N_t \) is the quantity of aggregate labor and \( N_t(h) \) is the input of labor of type \( h \), and where \( \theta_w \) denotes the elasticity of substitution between any two labor types. Aggregate labor \( N_t \) is then used as an input in the production of intermediate goods. Equilibrium in the labor market thus requires
\[ N_t = \int_0^1 L_t(f) \, df. \]
Here, it is important to notice the difference between \( L_t(f) \), the demand for aggregate labor emanating from firm \( f \), and \( N_t(h) \), the supply of labor of type \( h \) by the representative household.

The zero-profit condition yields the relation
\[ W_t = \left( \int_0^1 W_t(h)^{1-\theta_w} \, dh \right)^{1/(1-\theta_w)}, \]
where \( W_t \) is the nominal wage paid to aggregate labor while \( W_t(h) \) is the nominal wage paid to labor of type \( h \).

Mirroring prices, we assume that wages are subject to nominal rigidities, à la Calvo, in the manner of Erceg et al. (2000). Formally, unions face a constant probability \( \alpha_w \) of not being able to re-optimize wages. In the event that union \( h \) is not drawn to re-optimize at \( t \), it re-scales its wage according to the indexation rule
\[ W_t(h) = e^{\gamma_{z,t}[(\Pi)^{1-\iota_w}(\Pi_{t-1})^{\iota_w}]^{\gamma_w}W_{t-1}(h)} \]
where, as before, wages are indexed to a geometric average of steady-state inflation and past inflation, with \( \iota_w \in [0,1] \). However, we assume that the degree of indexation is here too less than perfect by imposing \( 0 \leq \gamma_w < 1 \). In addition, nominal wages are also indexed to average productivity growth with indexation degree \( 0 \leq \gamma_z < 1 \).

If drawn to re-optimize in period \( t \), a union chooses \( W_t^* \) in order to maximize
\[ \mathbb{E}_t \sum_{s=0}^{\infty} (\beta \alpha_w)^s \left\{ (1 + \tau_{w,t})\Lambda_{t+s} \frac{V_{t,t+s}^{\Pi}}{P_{t,s}^{\Pi}} W_t^* N_{t,t+s} - \frac{\chi}{\psi} N_{1,1,t+s}^{1+v} \right\} \]
where the demand function at $t+s$ facing a union who last revised its wage at $t$ obeys

$$N_{t,t+s} = \left( \frac{V_{t,t+s}^w}{W_{t+s}} \right)^{-\theta_w} N_{t+s}$$

and where $V_{t,t+s}^w$ reflects the compounded effects of wage indexation to past inflation and average productivity growth

$$V_{t,t+s}^w = e^{\gamma z \mu z (t+s)} \prod_{j=t}^{t+s-1} \left[ (\Pi)^{1-iw} (\Pi_j)^{iw} \right]^{\gamma w}.$$ 

Furthermore, we set $\tau_w$ so as to neutralize the steady-state distortion induced by wage markups.

### 2.4 Monetary Policy and the ZLB

Monetary policy in "normal times" is assumed to be given by a Taylor-like interest rate rule

$$\hat{i}_t = \rho \hat{i}_{t-1} + (1 - \rho) \left( a_\pi \hat{\pi}_t + a_y \hat{x}_t \right) + \zeta_{R,t}$$

(3)

where $i_t \equiv -\log(Q_t)$, with $\hat{i}_t$ denoting the associated deviation from steady state i.e., $\hat{i}_t \equiv i_t - i$. Also, $\pi_t \equiv \log \Pi_t$, $\hat{\pi}_t = \pi_t - \pi$ is the gap between inflation and its target, and $\hat{x}_t = \log(Y_t/Y^n_t)$ where $Y^n_t$ is the natural level of output, defined as the level of output that would prevail in an economy with flexible prices and wages and no cost-push shocks. Finally, $\zeta_{R,t}$ is a monetary policy shock.

Here, $\pi$ should be interpreted as the central bank target for change in the price index. An annual inflation target of 2% would thus imply $\pi = 2/400 = 0.005$ as the model will be parameterized and estimated with quarterly data. Note that the inflation target thus defined may differ from average inflation.

Crucially for our purpose, the nominal interest rate $i_t$ is subject to a ZLB constraint:

$$i_t \geq 0$$

The steady-state level of the real interest rate is defined by $r^* \equiv i - \pi$. Given logarithmic utility, it is related to technology and preference parameter according to $r^* = \rho + \mu z$. Combining these elements, it is convenient to write the ZLB constraint in terms the deviation of the nominal interest rate

$$\hat{i}_t \geq -(\mu z + \rho + \pi)$$

(4)

Thus, the rule effectively implemented is given by:

$$\hat{i}_t = \max \{ \rho \hat{i}_{t-1} + (1 - \rho) \left( a_\pi \hat{\pi}_t + a_y \hat{x}_t \right) + \zeta_{R,t}, -(\mu z + \rho + \pi) \}$$

Before proceeding, a remark is in order. The inflation target, $\pi$, is not assumed to be optimal. Note also that realized inflation might be on average below the target as a consequence of ZLB episodes, i.e. $E\{\pi_t\} < \pi$. In such instances of ZLB, monetary policy fails to deliver the appropriate degree of accommodation,
resulting in a more severe recession and lower inflation than in an economy in which there would not be a ZLB constraint.\footnote{For convenience, Table A.1 in the Appendix summarizes the various notions of optimal inflation and long-run or target inflation considered in this paper}

As equation (4) makes clear, $\mu_z$, $\rho$, $\pi$ have symmetric roles in the ZLB constraint. Put another way, for given structural parameters and a given process for $i_t$, the probability of hitting the ZLB would remain unchanged if productivity growth or the discount rate decline by one percent and the inflation target is increased by a commensurate amount at the same time. Based on these observations, one may be tempted to argue that in response to a permanent decline in $\mu_z$ or $\rho$, the optimal inflation target $\pi^*$ will change by the same amount (with a negative sign).

The previous conjecture is, however, incorrect. The reasons for this are twofold. First, any change in $\mu_z$ (or $\rho$) also translates into a change in the coefficients of the equilibrium dynamic system. It turns out that this effect is non-negligible since, as we show later, after a one percentage point decline in $r^*$ the inflation target has to be raised by more than one percent in order to keep the probability of hitting the ZLB unchanged. Second, because there are welfare costs associated with increasing the inflation target, the policy maker would also have to balance the benefits of keeping the incidence of ZLB episodes constant with the additional costs in terms of extra price dispersion and inefficient resource allocation. These costs can be substantial and more than compensate for the benefits of holding the probability of ZLB constant. Assessing these forces is precisely this paper’s endeavor.

2.5 Solution Method

Because the model has a stochastic trend, we first induce stationarity by dividing trending variables by $Z_t$. The resulting system is then log-linearized in the neighborhood of its deterministic steady state.\footnote{See the Technical Appendix for further details.} We append to the system a set of equations describing the dynamics of the structural shocks, namely

$$\zeta_{k,t} = \rho_k \zeta_{k,t-1} + \sigma_k e_{k,t}, \quad e_{k,t} \sim N(0,1)$$

for $k \in \{R, g, u, q, z\}$.

Absent the ZLB constraint, the model can be solved and cast into the usual linear transition and observation equations:

$$s_t = T(\theta)s_{t-1} + R(\theta)e_{t}, \quad x_t = M(\theta) + H(\theta)s_t,$$

with $s_t$ a vector collecting the model’s state variables, $x_t$ a vector of observable variables and $e_t$ a vector of innovations to the shock processes $e_t = (e_{R,t}, e_{g,t}, e_{u,t}, e_{q,t}, e_{z,t})'$. The solution coefficients are regrouped in the conformable matrices $T(\theta)$, $R(\theta)$, $M(\theta)$, and $H(\theta)$ which depend on the vector of structural parameters $\theta$. 


The model becomes non-linear when one allows the ZLB constraint to bind. The solution method we implement follows the approach developed by Bodenstein et al. (2009) and Guerrieri and Iacoviello (2015). The approach can be described as follows. There are two regimes: the no-ZLB regime \( k = n \) and the ZLB regime \( k = e \) and the canonical representation of the system in each regime is

\[
\mathbb{E}_t \{ A^{(k)} s_{t+1} + B^{(k)} s_t + C^{(k)} s_{t-1} + D^{(k)} \varepsilon_t \} + f^{(k)} = 0
\]

where \( A^{(k)}, B^{(k)}, C^{(k)}, \) and \( D^{(k)} \) are conformable matrices and \( f^{(k)} \) is a vector of constants. In the no-ZLB regime, the vector \( f^{(n)} \) is filled with zeros. In the ZLB regime, the row of \( f^{(e)} \) associated with \( i_t \) is equal to \( \mu_z + \rho + \pi \). Similarly, the rows of the system matrices associated with \( i_t \) in the no-ZLB regime correspond to the coefficients of the Taylor rule while in the ZLB regime, the coefficient associated with \( i_t \) is equal to 1 and all the other coefficients are set to zero.

In each period \( t \), given an initial state vector \( s_{t-1} \) and vector stochastic innovations \( \varepsilon_t \), we simulate the model under perfect foresight (i.e., assuming that no further shocks hit the economy) over the next \( N \) periods, for \( N \) sufficiently large. In case this particular draw is not conducive to a ZLB episode, we find \( s_t \) using the linear solution stated above. In contrast, if this draw leads to a ZLB episode, we postulate integers \( N_e < N \) and \( N_x < N \) such that the ZLB is reached at time \( t + N_e \) and left at time \( t + N_x \). In this case, we solve the model by backward induction. We obtain the time varying solution

\[
s_{t+q} = d_{t+q} + T_{t+q} s_{t+q-1} + R_{t+q} \varepsilon_{t+q}
\]

where, for \( q \in \{ N_e, \ldots, N_x - 1 \} \)

\[
T_{t+q} = -\left( A^{(e)} T_{t+q+1} + B^{(e)} \right)^{-1} C^{(e)}, \quad R_{t+q} = -\left( A^{(e)} T_{t+q+1} + B^{(e)} \right)^{-1} D^{(e)}
\]

and, for \( q \in \{ 0, \ldots, N_e - 1 \} \)

\[
T_{t+q} = -\left( A^{(n)} T_{t+q+1} + B^{(n)} \right)^{-1} C^{(n)}, \quad R_{t+q} = -\left( A^{(n)} T_{t+q+1} + B^{(n)} \right)^{-1} D^{(n)}
\]

using \( T_{t+N_e} = T, \quad R_{t+N_e} = R, \) and \( d_{t+N_e} \) set to a column filled with zeros as initial conditions of the backward recursion.

We then check that given the obtained solution, the system hits the ZLB at \( t + N_e \) and leaves the ZLB at \( t + N_x \). Otherwise, we shift \( N_e \) and/or \( N_x \) forward or backward by one period and start all over again until convergence. Once convergence has been reached, we use the resulting matrices to compute \( s_t \) and repeat the process for all the simulation periods.

Our approach is thus similar to the one used by Coibion et al. (2012) in their study of the optimal inflation target in a New Keynesian setup.\(^4\)

\(^4\)In practice we combine the implementation of the Bodenstein et al. (2009) algorithm developed by Coibion et al. (2012) with
2.6 Estimation Results

We estimate the model using data for a pre-crisis period over which the ZLB constraint is not binding. This enables us to use the linear version of the model. The sample of observable variables is \( X_T \equiv \{ x_t \}_{t=1}^T \) with

\[
x_t = [\Delta \log(GDP_t), \Delta \log(GDP \text{ Deflator}_t), \Delta \log(Wages_t), \text{Short Term Interest Rate}_t]'
\]

where the short term nominal interest rate is the effective Fed Funds Rate for the US and the Euribor 3 months rate for the Euro-Area. We use a sample of quarterly data covering the period 1985Q2-2009Q4. This choice is guided by two objectives. First, this sample strikes a balance between size and the concern of having a homogeneous monetary policy regime over the period considered. In the US case, the sample covers the Volcker and post-Volcker period, arguably one of relative homogeneity of monetary policy. For the euro area, the sample starts approximately when the disinflation policies were simultaneously conducted in the main euro area countries (see Fève et al. 2010) and then covers the single currency period. Here too, this corresponds to a period of relative monetary policy homogeneity. Second, we use a sample that coincides more or less with the so-called Great Moderation. Over the latter, as has been argued in the literature, we expect smaller shocks to hit the economy. In principle, this will lead to a conservative assessment of the effects of the more stringent ZLB constraint due to lower real interest rates.\(^5\)

The parameters \( \phi, \theta_p, \theta_w, \iota_p, \) and \( \iota_w \) are calibrated prior to estimation. Given our specification of the measurement equation, the parameters \( \iota_p \) and \( \gamma_p \) and, likewise, \( \iota_w \) and \( \gamma_w \), are not separately identified. In effect, they appear in the dynamic system as products \( \gamma_p \iota_p \) or \( \gamma_w \iota_w \). We thus set \( \iota_p = \iota_w = 1 \) at the calibration stage. As long as \( \iota_w \) and \( \iota_w \) are not set to zero, the particular value chosen has no other consequence on the estimation results than a reinterpretation of \( \gamma_w \) and \( \gamma_p \). The parameter \( \phi \) is set to \( 1/0.7 \), resulting in a steady-state labor share of 70%. The parameter \( \theta_p \) is set to 6, resulting in a steady-state markup of 20%. Similarly, the parameter \( \theta_w \) is set to 3, resulting in a wage markup of 50%.

We rely on a full-system Bayesian estimation approach to estimate the other model parameters. After having cast the dynamic system in the state-space representation for the set of observable variables, we use the Kalman filter to measure the likelihood of the observed variables. We then form the joint posterior distribution of the structural parameters by combining the likelihood function \( p(X_T|\theta) \) with a joint density characterizing some prior beliefs \( p(\theta) \). The joint posterior distribution thus obeys

\[
p(\theta|X_T) \propto p(X_T|\theta)p(\theta),
\]

Given the specification of the model, the joint posterior distribution cannot be recovered analytically but may be computed numerically, using a Monte-Carlo Markov Chain (MCMC) sampling approach. More

5The data are obtained from the Fred database for the US and from the “Area Wide Model” database of Fagan et al. (2001) and Eurostat national accounts for the Euro Area. In both cases, the GDP is expressed in per capita terms.

the solution algorithm and the parser from Dynare. Our implementation is in the spirit of Guerrieri and Iacoviello (2015), resulting in a less user-friendly yet faster suite of programs.
### Table 1: Estimation Results - US

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Prior Shape</th>
<th>Prior Mean</th>
<th>Prior std</th>
<th>Post. Mean</th>
<th>Post. std</th>
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<th>High</th>
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Note: ‘std’ stands for Standard Deviation, ‘Post.’ stands for Posterior, and ‘Low’ and ‘High’ denote the bounds of the 90% probability interval for the posterior distribution.

Specifically, we rely on the Metropolis-Hastings algorithm to obtain a random draw of size 1,000,000 from the joint posterior distribution of the parameters.

Tables 1 and 2 present the parameter’s postulated priors (type of distribution, mean, and standard error) and estimation results, i.e., the posterior mean and standard deviation, together with the bounds of the 90% probability interval for each parameter.

For the parameters $\pi$, $\mu_z$ and $\rho$, we impose Gaussian prior distributions. The parameters governing the latter are chosen so as to match the mean values of inflation, GDP growth, and the real interest rate in our US and euro area samples. Other than for these three parameters, we use the same prior distributions for the structural parameters in both the US and the euro area. Our choice of priors are standard. In particular, we use Beta distributions for parameters in $[0,1]$, Gamma distributions for positive parameters, and Inverse Gamma distributions for the standard error of the structural shocks.

The estimation results suggest several key differences between the US and the euro area.

First, consistent with the sample period, we find a higher growth rate $\mu_z$ in the euro area than in the US. Given that both economies have similar discount rates, this will result in a higher steady-state real interest rate in the euro area than in the US. This difference will play an important role later when we assess (i) the level of the optimal inflation target and (ii) the effects of a lower steady-state real interest rate. Second, we find generically higher degrees of indexation to past inflation in the US than in the euro area. This will translate into a higher tolerance for inflation in the US in our subsequent analysis of the optimal inflation...
### Table 2: Estimation Results - EA

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</table>

**Note:** ‘std’ stands for Standard Deviation, ‘Post.’ stands for Posterior, and ‘Low’ and ‘High’ denote the bounds of the 90% probability interval for the posterior distribution.

Target. This is because a higher indexation helps to mitigate the distortions induced by a higher inflation target. Everything else equal, we thus expect a higher optimal inflation target in the US than in the euro area. Third, we obtain broadly similar parameters for the shocks processes. One exception, though, is the so-called risk-premium shock. The unconditional variance of the US shock is somewhat higher than its euro area counterpart.

For the US, most of our estimated parameters are in line with the calibration adopted by Coibion et al. (2012), with important qualifications. First, we obtain a slightly higher degree of price rigidity than theirs (0.67 versus 0.55). Second, our specification of monetary policy is different from theirs. In particular, they allow for two lags of the nominal interest rate in the monetary policy rule while we only have one lag. However, we can compare the overall degree of interest rate smoothing in the two setups. To this end, abstracting from the other elements of the rule, we simply focus on the sum of autoregressive coefficients. It amounts to 0.92 in their calibration while the degree of smoothing in our setup has a mean posterior value of 0.85. While this might not seem to be a striking difference, it is useful to cast these figures in terms of half-life of convergence in the context of autoregressive model of order 1. Our value implies twice as small a half-life than theirs. Third, our monetary policy shock and our shocks to demand have approximately twice as small an unconditional standard deviation as theirs.
3 The Optimal Inflation Target

A second-order approximation of the household expected utility derived from the structural model is used to quantify welfare, in a similar manner as in Woodford (2003). Let $\mathcal{W}(\pi; \theta)$ denote this welfare criterion. This notation emphasizes that welfare depends on the inflation target $\pi$ together with the rest of the structural parameters $\theta$. Two cases are considered concerning the latter. In the baseline case the structural parameters $\theta$ are fixed at reference values and taken to be known with certainty by the policy maker. In an alternative exercise, the policy maker maximizes welfare while recognizing the uncertainty associated with the model’s parameters.

3.1 Baseline

The optimal inflation target associated with a given vector of parameters $\theta$, $\pi^*(\theta)$ is approximated via numerical simulations of the model allowing for an occasionally binding ZLB constraint, using the algorithm outlined above. The optimal inflation rate associated to a particular vector of parameters $\theta$ is then obtained as the one maximizing the welfare function, that is:

$$\pi^*(\theta) \equiv \arg \max_{\pi} \mathcal{W}(\pi; \theta).$$

For illustrative purpose, figures 1a and 1b display the welfare function for the US and the euro area – expressed as losses relative to the maximum social welfare – associated with three natural benchmarks for the parameter vector $\theta$: the posterior mean (dark blue line), the median (light blue line), and the mode (lighter blue line). For convenience, the peak of each welfare function is identified with a dot of the same color. Also, to facilitate interpretations, the inflation targets are expressed in annualized percentage rates.

As Figure 1a illustrates, the US optimal inflation target is close to 2% and varies between 1.85% and 2.21% depending on which indicator of central tendency (mean, mode, or median) is selected. This range of values is consistent with the ones of Coibion et al. (2012) even though in the present paper it is derived from an estimated model over a much shorter sample. Importantly, the high degree of interest rate smoothing obtained in Coibion et al. (2012) largely compensates for their larger shocks. In the euro area, as figure 1b reports, the optimal inflation target is close to 1.5% and varies between 1.31% and 1.58%. Altogether, these

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6See the Technical Appendix for details.

7More precisely, a sample of size $T = 100000$ of innovations $\{\epsilon_t\}_{t=1}^T$ is drawn from a Gaussian distribution (we also allow for a burn in sample of 200 points that we later discard). We use these shocks to simulate the model for given parameter vector $\theta$. The welfare function $\mathcal{W}(\pi; \theta)$ is approximated by replacing expectations with sample averages. The procedure is repeated for each of $K = 51$ inflation targets on the grid $\{\pi^{(k)}\}_{k=1}^K$ ranging from $\pi = 0.5/4\%$ to $\pi = 5/4\%$ (expressed in quarterly rates). Importantly, we use the exact same sequence of shocks $\{\epsilon_t\}_{t=1}^T$ in each and every simulation over the inflation grid.

8Coibion et al. (2012) calibrate their model on a post-WWII, pre-Great Recession US sample. By contrast, we use a Great Moderation sample.
Figure 1: Examples of loss functions

(a) US

(b) EA

Note: Blue: parameters set at the posterior mean; light blue: parameters set at the posterior median; Lighter blue: parameters set at the posterior mode. \( \pi^* \equiv \text{log}(\Pi^*) \). In all cases, the welfare functions are normalized so as to peak at 0.

numbers seem roughly consistent with the quantitative inflation targets adopted by the Fed and the ECB, respectively.

To complement on these illustrative results, figures 2a and 2b display the probability of reaching the ZLB as a function of the annualized inflation target (again, with the parameter vector \( \theta \) evaluated at the posterior mean, median, and mode). For convenience, the circles in each curve mark the corresponding optimal inflation target.

The probabilities of reaching the ZLB are relatively low, at about 6% for the US and about 10% for the euro area. This result, as anticipated above, is the mere reflection of our choice of a Great Moderation sample.

3.2 Accounting for Parameter Uncertainty

An interesting feature of the welfare functions reported in figures 1a and 1b is that they are strongly asymmetric: adopting an inflation target 1 percentage point below the optimal value generates welfare losses larger than setting it 1 percentage point above. Moreover, as 1a and 1b suggest, the location of the loss function also depends on the parameter considered. This is important since the exact values of \( \theta \) and \( \pi^*(\theta) \) are subject to uncertainty. As a result, a policy maker that seeks to maximize expected welfare while recognizing the uncertainty surrounding \( \theta \) will tend to choose a relatively higher inflation target compared to the case where \( \theta \) is taken to be known with certainty, as in the baseline analysis above.

Formally, the estimated posterior distribution of parameters \( p(\theta|X_T) \) can be exploited to quantify the
impact of such parameter uncertainty on the optimal inflation target and to compute a "Bayesian-theoretic optimal inflation target". We define the latter as the inflation target $\pi^{**}$ which maximizes the expected welfare not only over the realizations of shocks but also over the realizations of parameters\footnote{This Bayesian inflation target is recovered from simulating the model under a ZLB constraint using the exact same sequence of shocks $\{\epsilon_t\}_{t=1}^T$ with $T = 100000$ as in the previous subsection (together with the same burn-in sample) and combining it with $N$ draws of parameters $\{\theta_j\}_{j=1}^N$ from the estimated posterior distribution $p(\theta|X_T)$, with $N = 500$. As in the previous section, the social welfare function $W(\pi; \theta)$ is evaluated for each draw of $\theta$ over a grid inflation targets $\{\pi(k)\}_{k=1}^K$. The Bayesian welfare criterion is then computed as the average welfare across parameter draws. Here, we start with the same inflation grid as before and then run several passes. In the first pass, we identify the inflation target maximizing the Bayesian welfare criterion. We then set a finer grid of $K = 51$ inflation targets around this value. We repeat this process several times with successively finer grids of inflation targets until the identified optimal inflation target proves insensitive to the grid.}

$$\pi^{**} \equiv \arg \max_{\pi} \int_\theta W(\pi; \theta) p(\theta|X_T) d\theta.$$ 

We interpret the spread between the inflation target at the posterior mean $\bar{\theta}$ and the Bayesian inflation target as a measure of how uncertainty about the parameter value could be conducive to a larger inflation buffer to hedge against particularly detrimental parameter values (either because they lead to more frequent ZLB episodes or because they lead to particularly acute inflation distortions). We define

$$\text{Spr}(\theta) \equiv \pi^{**} - \pi^*(\theta)$$

and assess below $\text{Spr}(\bar{\theta})$.

According to the simulation exercise, $\pi^{**} = 2.40\%$ for the US and $\pi^{**} = 2.20\%$ for the euro area. In both cases, these robust optimal inflation targets are larger than the values obtained with $\theta$ set at its
central tendency. As expected, a Bayesian policy maker choose a higher inflation target to hedge against particularly harmful states of the world (i.e., parameter draws) where the frequency of hitting the ZLB is high. Noticeably, this inflation cushion is substantially higher in the euro area where Spr(\(\bar{\theta}\)) = 0.62%, while in the US, Spr(\(\bar{\theta}\)) = 0.19%. This occurs in spite of the optimal inflation target being lower in the euro area than in the US when evaluated at the posterior mean of the parameter vector \(\bar{\theta}\).\(^{10}\)

4 The Optimal Inflation Target and the Steady State Real Interest Rate

The focus of this section is to investigate how the monetary authority should adjust its optimal inflation target \(\pi^*\) in response to changes in the steady-state real interest rate, \(r^*\). Intuitively, with a lower \(r^*\) the ZLB is bound to bind more often, so one would expect a higher inflation target should be desirable in that case. But the answer to the practical question of by how much should the target be increased is not obvious. Indeed, the benefit of providing a better hedge against hitting the ZLB, which is an infrequent event, comes at a cost of higher steady-state inflation which induces permanent costs, as argued by, e.g., Bernanke (2016).

To start with, we compute the relation linking the optimal inflation target to the steady-state real interest rate, based on simulations of the model and ignoring parameter uncertainty. We show that the link between \(\pi^*\) and \(r^*\) depends to some extent on the reason underlying a variation in \(r^*\), i.e. a change in the discount rate \(\rho\) or a change in growth rate of technology \(\mu_z\). We also investigate the role of parameter uncertainty and, in particular, uncertainty about \(r^*\), in the determination of the Bayesian-theoretic optimal inflation target. Finally, we investigate how the relation between the optimal inflation target and the steady-state real interest rate depends on whether the central bank (i) targets the average of realized inflation rather than (non-stochastic) steady-state inflation rate, (ii) operates in an economy where the lower bound on the nominal interest rate is negative or (iii) knows its policy reaction function with certainty.

4.1 The baseline \((r^*, \pi^*)\) relation

To characterize the link between \(r^*\) and \(\pi^*\), the following simulation exercise is conducted. The structural parameter vector \(\theta\) is fixed at its posterior mean, \(\bar{\theta}\), with the exception of \(\mu_z\) and \(\rho\). These two parameters are varied – each in turn, keeping the other parameter, \(\mu_z\) or \(\rho\), fixed at its baseline posterior mean value. For both \(\mu_z\) and \(\rho\), we consider values on a grid ranging from 0.4% to 10% in annualized percentage terms. The model is then simulated for each possible values of \(\mu_z\) or \(\rho\) and various values of inflation targets \(\pi\) using the procedure as before.\(^{11}\) The optimal value \(\pi^*\) associated to each value of \(r^*\) is obtained as the one

\(^{10}\)Figures B.1a and B.1b in Appendix show the posterior distribution of \(\pi^*(\bar{\theta})\). It is broadly symmetric in the US and shows substantial asymmetry in the euro area.

\(^{11}\)In particular, we use the same sequence of shocks \(\{e_t\}_{t=1}^T\) as used in the computation implemented in the baseline exercises of Section 3. Here again, we start from the same grid of inflation targets for all the possible values of \(\mu_z\) or \(\rho\). Then, for each value
maximizing the welfare criterion $\mathcal{W}(\pi; \theta)$.\textsuperscript{12}

We finally obtain two curves. The first one links the optimal inflation target $\pi^*$ to the steady-state real interest rate $r^*$ for various growth rate of technology $\mu_z$: $\pi^*(r^*(\mu_z))$, where the notation $r^*(\mu_z)$ highlights that the steady-state real interest rate varies as $\mu_z$ varies. The second one links the optimal inflation target $\pi^*$ to the steady-state real interest rate $r^*$ for various discount rates $\rho$: $\pi^*(r^*(\rho))$. Here, the notation $r^*(\rho)$ highlights that the steady-state real interest rate varies as $\rho$ varies.\textsuperscript{13}

Figures 3a and 3b depict the $(r^*, \pi^*)$ relations thus obtained for the US and the euro area, respectively. The blue dots correspond to the case when the real steady-state interest rate $r^*$ varies with $\mu_z$. The red dots correspond to the case when the real steady-state interest rate $r^*$ varies with $\rho$. For convenience, both the real interest rate and the associated optimal inflation target are expressed in annualized percentage rates. The dashed grey lines indicate the benchmark result corresponding to the optimal inflation target at the posterior mean of the structural parameter distribution. To complement these results, Figures 4a and 4b show the relation between $r^*$ and the probability of hitting the ZLB, \textit{evaluated at the optimal inflation target}, for the US and the euro area, respectively. As before, the blue dots correspond to the case when the real steady-state interest rate $r^*$ varies with $\mu_z$. The red dots correspond to the case when the real steady-state $\mu_z$ or $\rho$, we refine the inflation grid over successive passes until the optimal inflation target associated with a particular value of $\mu_z$ or $\rho$ proves insensitive to the grid.

\textsuperscript{12}To illustrate the construction of this figure, see Appendix C. There, we show how two particular points of this curve are derived from the welfare criteria.

\textsuperscript{13}Figures D.1a, D.1b, D.2a, and D.2b report similar results at the posterior mode and at the posterior median.
As expected the relations in 3a and 3b are decreasing. However, the slope varies with the value of $r^*$. For both the US and the euro area, the slope is relatively large in absolute value – although smaller than one – for moderate values of $r^*$ (say below 4 percent). The slope declines in absolute value as $r^*$ increases: Lowering the inflation target to compensate for an increase in $r^*$ becomes less and less desirable. This reflects the fact that, as $r^*$ increases, the probability of hitting the ZLB becomes smaller and smaller. For very large $r^*$ values, the probability becomes almost zero, as Figures 4a and 4b show.

At some point, the optimal inflation target becomes insensitive to changes in $r^*$ when the latter originate from changes in the discount rate $\rho$. In this case, the inflation target stabilizes at a slightly negative value, in order to lower the nominal wage inflation rate required to support positive productivity growth, given the imperfect indexation of nominal wages to productivity. At the steady state, the real wage must grow at a rate of $\mu_z$. It is optimal to obtain this steady-state growth as the result of a moderate nominal wage increase and a moderate price decrease, rather than the result of a zero price inflation and a consequently larger nominal wage inflation.\(^\text{15}\)

\(^{14}\)Figures E.1a and E.1b in Appendix show the relation between $r^*$ and the nominal interest rate when the inflation target is set at its optimal value.

\(^{15}\)For very large $r^*$, as a rough approximation, we can ignore the effects of shocks and assume that the ZLB is a zero-mass event. Assuming also a negligible difference between steady-state and efficient outputs and letting $\lambda_p$ and $\lambda_w$ denote the weights attached to price dispersion and wage dispersion, respectively, in the approximated welfare function, the optimal inflation obeys $\pi^* \approx -\lambda_w(1 - \gamma_z)(1 - \gamma_w)/[\lambda_p(1 - \gamma_p)^2 + \lambda_w(1 - \gamma_w)^2]\mu_z$. Given the low values of $\lambda_w$ resulting from our estimation, it is not surprising that $\pi^*$ is negative but close to zero. See Amano et al. (2009) for a similar point in the context of a model abstracting...
The previous tension is even more apparent when \( r^* \) varies with \( \mu_z \) since, in this case, the effects of imperfect indexation of wages to productivity are magnified given that a higher \( \mu_z \) calls for a higher growth in the real wage, which is optimally attained through greater price deflation, as well as a higher wage inflation. Notice however that even in this case, the optimal inflation target becomes little sensitive to changes in \( r^* \) for very large values of \( r^* \), typically above 6%, both in the US and the euro area.

For low values of \( r^* \), on the other hand, the slope of the curve is steeper. In particular, in the empirically relevant region, the relation is not far from one-to-one. More precisely, it shows that, starting from the posterior mean estimate of \( \theta \), a 100 basis points decline in \( r^* \) should lead to a +99 basis points increase in \( \pi^* \) in the US and to a +81 basis points increase in the euro area. Importantly, this increase in the optimal inflation target is the same no matter the underlying factor causing the change in \( r^* \): a drop in potential growth, \( \mu_z \), or a decrease in the discount factor, \( \rho \). At the same time, the probability of ZLB evaluated at the optimal inflation rate also increases when the real rate decreases. In the US case, at some point, the speed at which this probability increases slows down, reflecting that the social planner would choose to increase the inflation target to almost compensate for the higher incidence of ZLB episodes. By contrast, in the euro area, the incidence of ZLB seems to increase substantially after a decline in the real interest rate, even at low values of the latter.

To gain insight into this striking difference, Figures 5a and 5b show how the probability of ZLB changes as a function of \( r^* \), holding the inflation target constant. We first set the inflation target at its optimal from ZLB issues.
baseline value (i.e., the value computed at the posterior mean, 2.21 for the US and 1.58 for the euro area). This is reported below as the blue dots. Similarly, we also compute an analog relation assuming this time that the inflation target is held constant at the optimal value consistent with a steady-state real interest rate one percentage point lower (thus, inflation is set to 3.20 for the US and 2.39 for the euro area). Here again, the other parameters are set at their posterior mean. This corresponds to the red dots in the figure.

Consider first the blue line. At the level of the real interest rate prevailing before the permanent decline, assuming that the Central Bank sets its target to the associated optimal level, the probability of reaching the ZLB would be slightly below 6% in the US and close to 9% in the euro area. Imagine now that the real interest rates experiences a decline of 100 basis points. Keeping the inflation target at the same level as prior to the shock, the probability of reaching the ZLB would now climb up to approximately 11% in the US and 16% in the euro area. However, the change in the optimal inflation target brings the probability of reaching the ZLB back to approximately 6% in the US and 11% in the euro area. In the euro area, the social planner is willing to tolerate a smaller inflation target than the one that would fully neutralize the effects of the natural rate decline on the probability of hitting the ZLB. By way of contrast, the social planner in the US would almost neutralize this effect. In this sense, the US economy has a greater tolerance for steady-state inflation than the euro area. This is in part a consequence of the different estimates for the degree of indexation of prices to past inflation found at the estimation stage.

4.2 Accounting for Parameter Uncertainty

Next we investigate the impact of parameter uncertainty on the relation between the optimal inflation target and the steady-state real interest rate. Specifically, we want to determine how the Bayesian-theoretic optimal inflation target $\pi^{\star\star}$ reacts to a downward shift in the distribution of the steady-state real interest rate $r^\star$.

Assessing how such a change affects $\pi^{\star\star}$ for every value of $r^\star$ is not possible due to the computational cost involved. Such a reaction is thus investigated for a particular scenario: it is assumed that the economy starts from the posterior distribution of parameters $p(\theta|X_T)$ and that, everything else being constant, the mean of $r^\star$ decreases by 100 basis points. Such a 1 percentage point decline is chosen mainly for illustrative purposes. Yet, it is of a comparable order of magnitude, although relatively smaller in absolute value, as recent estimates of the drop of the natural rate after the crisis such as Laubach and Williams (2016) and Holston et al. (2017). The counterfactual exercise considered can therefore be seen as a relatively conservative characterization of the shift in steady-state real interest rate. Figures 6a and 6b depict the counterfactual shift in the distribution of $r^\star$ that is considered for, respectively, the US and the euro area.

The Bayesian-theoretic optimal inflation target corresponding to the counterfactual lower distribution of $r^\star$ is obtained from a simulation exercise that relies on the same procedure as before.\footnote{Again, we use the same sequence of shocks and the same parameter draws as in section 3.} Given a draw in
the posterior of parameter vector $\theta$, the value of the steady-state real interest rate is computed using the expression implied by the postulated structural model $r^*(\theta) = \rho(\theta) + \mu_z(\theta)$. From this particular draw, a counterfactual lower steady-state real interest rate, $r^*(\theta\Delta)$, is obtained by shifting the long-run growth component of the model $\mu_z$ downwards by 1 percentage point (in annualised terms). The welfare function $W(\pi; \theta\Delta)$ is then evaluated. Since there are no other changes than this shift in the mean value of $\mu_z$ in the distribution of the structural parameters, we can characterize the counterfactual distribution $p(\theta\Delta|X_T)$ as a simple transformation of the estimated posterior $p(\theta|X_T)$. The counterfactual Bayesian-theoretic optimal inflation target is then obtained as

$$\pi^{**}_\Delta \equiv \arg\max_\pi \int_{\theta_\Delta} W(\pi; \theta_\Delta) p(\theta_\Delta|X_T) d\theta_\Delta.$$  

Figures 7a and 7b illustrate the counterfactual change in optimal inflation target obtained when the steady-state real interest rate declines by 100 basis points and its new value stays uncertain. For the US, the simulation exercise returns a value of $\pi^{**}_\Delta = 3.30\%$ i.e. 90 basis points higher than the optimal value under uncertainty obtained with the posterior distribution of parameters obtained on a pre-crisis sample $\pi^{*}_\Delta = 2.40\%$. For the euro area, $\pi^{**}_\Delta = 3.10\%$, also 90 basis points higher than the optimal value $\pi^{**}_\Delta = 2.20\%$ obtained with the baseline posterior distribution of parameters.\footnote{Figures F.1a and F.1b in Appendix show how the posterior distribution of $\pi^*$ is shifted after the permanent decline in the mean of $r^*$.}

Thus, we see that a monetary authority that is concerned about the uncertainty surrounding the parameters driving the costs and benefits of the inflation target raises the optimal inflation target but does not
alter the reaction of this optimal inflation target following a drop in $r^*$: in both cases, a 100 basis points decrease in the steady-state real interest rate calls for a roughly 90 basis point increase in the optimal inflation target in the vicinity of pre-crisis parameter estimates.

4.3 Further Experiments

In the present section we carry out four additional exercises related to the optimal adjustment of the inflation target in response to a change in the steady-state real interest rate. The first three exercises examine the implications of three alternative assumption regarding monetary policy. The fourth exercise looks at the case of large shocks.

**Average vs Target Inflation** As emphasized in recent works (see, notably, Hills et al. 2016, Kiley and Roberts 2017), when the probability of hitting the ZLB is non-negligible, realized inflation is on average significantly lower than the inflation rate that the central bank targets in the interest rate rule (and which would correspond to steady-state inflation in the absence of shocks). This results from the fact that anytime the ZLB is binding (which happens recurrently) the central bank effectively loses its ability to stabilize inflation around the target. Knowing this, the central banks may want to define (and communicate) its target in terms of the effective average realized inflation. In this section, we investigate whether such an adjustment in the communication policy is warranted when the inflation target is chosen optimally.

To this end, the analysis of the $(r^*, \pi^*)$ relation of section 4.1 is complemented here with the analysis
of the relation between $r^*$ and the average realized inflation rate $\mathbb{E}\{\pi_t\}$ obtained when simulating the model for various values of $r^*$ and the associated optimal inflation target $\pi^*$. In the interest of brevity, the calculations are undertaken assuming that changes in average productivity growth $\mu_z$ is the only source of variation in the natural interest rate.

Figures 8a and 8b illustrate the difference between the $(r^*, \pi^*)$ curve (blue dots) and the $(r^*, \mathbb{E}\{\pi_t\})$ curve (red dots) for the US and the euro area. The overall shape of the curve is unchanged. Unsurprisingly, both curves are identical when $r^*$ is high enough. In this case, the ZLB is (almost) not binding and average realized inflation does not differ much from $\pi^*$. A spread between the two emerges for very low values of $r^*$. There, for low values of the natural rate, the ZLB incidence is higher and, as a result, average realized inflation becomes indeed lower than the optimal inflation target. However, that spread remains limited, less than 10 basis points. The reason is that the implied optimal inflation target is sufficiently high to prevent the ZLB from binding too frequently, thus limiting the extent to which average realized inflation and $\pi^*$ can differ.

Unreported simulation results show that the gap between $\pi^*$ and average realized inflation becomes more substantial when the inflation target is below its optimal value. For instance, mean inflation is roughly zero when the central bank adopts a 1% inflation target in an economy where the optimal inflation target is $\pi^* = 2\%$.

**A Negative Effective Lower Bound** The recent experience of many advanced economies (including the euro area) points to an effective lower bound (ELB) for the nominal interest rate below zero. For instance, the ECB’s deposit facility rate, which gears the overnight money market rate because of excess liquidity,
was set at a negative value of $-10$ basis points in June 2014 and has been further lowered down to $-40$ basis points in March 2016.

We use the estimated euro-area model to evaluate the implications of a negative ELB. More precisely, we set the lower bound on the nominal rate $i_t$ so that

$$i_t \geq e$$

and we set $e$ to $-40$ basis points (in annual terms) instead of zero. Results are presented in Figure 9. As expected, the $(r^*, \pi^*)$ locus is shifted downwards, though by somewhat less than 40 basis points. Importantly, its slope remains identical to the baseline case: a 100 basis points downward shift in the distribution of $r^*$ calls for a 90 basis points increase in $\pi^*$.

**A Known Reaction Function** Here we study the consequences of the (plausible) assumption that the central bank actually knows the coefficients of its interest rate rule with certainty. More specifically we repeat the same simulation exercise as in subsection 4.2 but with parameters $a_\pi$, $a_y$ and $\rho_i$ in the reaction function 3 taken to be known with certainty. In practice we fix these three parameters at their posterior mean, instead of sampling them from their posterior distribution. This is arguably the relevant approach from the point of view of the policymaker.$^{18}$ Note, however, that all the other parameters are subject to uncertainty from the stand-point of the central bank.

Figures 10a and 10b present, respectively for the US and the euro area, the Bayesian-theoretic optimal inflation targets obtained when simulating the model at the initial posteriors and after a -100 basis points shift.

---

$^{18}$In practice, long-run inflation targets are seldom reconsidered while the rotation in monetary policy committees happens at a higher frequency. From this viewpoint, our baseline assumption of uncertainty on all the monetary policy rule parameters is not necessarily unwarranted.
level shift in the posterior distribution of the long-run growth rate $\mu_z$ and, hence, the steady-state real rate $r^*$. According to these simulations, the inflation target should initially be $\pi^{**} = 2.24\%$ in the US and $\pi^{**} = 2.36\%$ in the euro area. After the counterfactual change in the distribution of $r^*$ considered, $\pi^{**}$ should be increased to 3.16% in the US and to 3.28% in the euro area, again in the ballpark of a 90 basis points increase in $\pi^*$ to compensate for the higher probability to hit the ZLB induced by a 100 basis points downward shift in the distribution of $r^*$.

**What if shocks are larger?** As argued before, the model is estimated using data from the Great Moderation period. One may legitimately argue that the decline in the real interest rate resulting from the secular stagnation has come hand in hand with larger shocks, as the Great Recession suggests. To address this concern, we simulate the model assuming that demand shocks have a 30% larger standard error as estimated.

We conduct this exercise assuming that changes in average productivity growth $\mu_z$ are the only driver of changes in the natural rate. Apart from $\sigma_q$ and $\sigma_y$, which are re-scaled, all the other parameters are frozen at their posterior mean. Given this setup, the optimal inflation target is 3.7% in the US and 2.7% in the euro area, as opposed to 2.21% and 1.58% in the baseline, respectively. Also, under the alternative shock configuration, the probability of hitting the ZLB is 5.3% in the US and 10.1% in the euro, as opposed to 5.5% and 9.4% in the baseline, respectively.

Figures 11a and 11b report the $(r^*, \pi^*)$ relation under larger demand shocks (red dots) and compares the

![Figure 10: $E_\theta(W(\pi, \theta))$](image)
outcome with what obtained in the baseline (blues dots).\textsuperscript{19} Interestingly, the $(r^*, \pi^*)$ locus has essentially the same slope in the low $r^*$ region. Here again, we find a slope close to -0.9. However, the curve is somewhat steeper in the high $r^*$ region and shifted up, compared to the baseline scenario. This reflects that under larger demand shocks, even at very high levels of the natural rate, a drop in the latter is conducive to more frequent ZLB episodes. The social planner is then willing to increase the inflation target at a higher pace than in the baseline scenario and generically sets the inflation target at higher levels to hedge the economy against ZLB episodes.

5 Summary and conclusion

In this paper, we assessed how changes in the steady-state natural interest rate translate into changes in the optimal inflation target in a model subject to the ZLB. Our main finding is that, starting from pre-crisis values, a 1% decline in the natural rate should be accommodated by an increase in the optimal inflation target of about 0.9%. For convenience, Table 3 recaps our results. Overall, across the different concepts of optimal inflation considered in this paper, the level of optimal inflation does vary. However it is a very robust finding that the slope of the $(r^*, \pi^*)$ relation is close to -0.9 in the vicinity of the pre-crisis value of steady-state real interest rates both in the US and in the euro-area.

In our analysis, we considered adjusting the inflation target as the only option at the policymaker’s dis-

\textsuperscript{19}We obtain this figure using the same procedure as outlined before. Here again, we run several passes with successively refined inflation grids.
Table 3: Effect of a decline in $r^*$ under alternative notions of optimal inflation

<table>
<thead>
<tr>
<th></th>
<th>US Baseline</th>
<th>Lower $r^*$</th>
<th>EA Baseline</th>
<th>Lower $r^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean of $\pi^*$</td>
<td>2.00</td>
<td>3.00</td>
<td>1.79</td>
<td>2.60</td>
</tr>
<tr>
<td>Median of $\pi^*$</td>
<td>1.96</td>
<td>2.90</td>
<td>1.47</td>
<td>2.28</td>
</tr>
<tr>
<td>$\pi^*$ at post. mean</td>
<td>2.21</td>
<td>3.20</td>
<td>1.58</td>
<td>2.39</td>
</tr>
<tr>
<td>$\pi^*$ at post. median</td>
<td>2.12</td>
<td>3.11</td>
<td>1.49</td>
<td>2.30</td>
</tr>
<tr>
<td>$\pi^{**}$, frozen MP</td>
<td>2.40</td>
<td>3.30</td>
<td>2.20</td>
<td>3.10</td>
</tr>
<tr>
<td>$\pi^*$ at post. mean, ELB -40 bp</td>
<td>—</td>
<td>—</td>
<td>1.31</td>
<td>2.08</td>
</tr>
<tr>
<td>Average realized inflation at post. mean</td>
<td>2.20</td>
<td>3.19</td>
<td>1.56</td>
<td>2.36</td>
</tr>
<tr>
<td>Average realized inflation at post. mean, ELB -40 bp</td>
<td>—</td>
<td>—</td>
<td>1.24</td>
<td>1.97</td>
</tr>
</tbody>
</table>

Note: all figures are in annualized percentage rate.

The complete comparison of these policy trade-offs goes beyond the scope of the present paper.

We have discussed the impact of higher inflation target, abstracting from the transition to a higher inflation target. In the current lowflation environment, increasing the inflation target in reaction to a drop in the steady-state value of the real interest rate might be challenging: because of more frequent ZLB episodes, the realizations of inflation might be on average below the initial inflation target for some time and increasing the inflation target therefore would raise some credibility issues.

Finally, our analysis also abstracted from forces identified in the literature as warranting a small, positive inflation target, irrespective of ZLB issues, as emphasized in Bernanke et al. (1999). The first is grounded on measurement issues, following the finding from the 1996 Boskin report that the consumer price index did probably over estimate inflation in the US by over 1 percentage point in the early nineties. The second argument is rooted in downward nominal rigidities. In an economy where there are such
downward rigidities (e.g. in nominal wages) a positive inflation rate can help "grease the wheel" of the labor market by facilitating relative price adjustments. Symmetrically, we also abstracted from forces calling for lower inflation targets. The most obvious is the so-called Friedman (1969) rule, according to which average inflation should equal to minus the steady state real interest rate, hence be negative, in order to minimize loss of resources or utility and the distortionary wedge between cash and credit goods (e.g. consumption and leisure) induced by a non-zero nominal interest rate. We conjecture that adding these elements to our setup would leave our main conclusions unchanged. A complete assessment is left for future research.
A Various long-run and optimal inflation rates considered

Table A.1: Various Notions of Long-run and optimal Inflation in the model

<table>
<thead>
<tr>
<th>Notion</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\pi$</td>
<td>Any inflation target, used to define the “inflation gap” that enters the Taylor rule</td>
</tr>
<tr>
<td>$E(\pi_t)$</td>
<td>Average realized inflation, might differ from $\pi$ due to ZLB</td>
</tr>
<tr>
<td>$\pi^*(\theta)$</td>
<td>Inflation target that minimizes the loss function given a structural parameters $\theta$</td>
</tr>
<tr>
<td>$\pi^*(\bar{\theta})$</td>
<td>$\pi^*$ assuming parameters at post. mean</td>
</tr>
<tr>
<td>$\pi^*(\text{median}(\theta))$</td>
<td>$\pi^*$ assuming parameters at post. median</td>
</tr>
<tr>
<td>$\bar{\pi}$</td>
<td>average of $\pi^<em>(\theta)$ over the posterior distribution of $\theta$, i.e., $\int_\theta \pi^</em>(\theta)p(\theta</td>
</tr>
<tr>
<td>Median($\pi^*$)</td>
<td>Median of $\pi^*(\theta)$ over the posterior distribution</td>
</tr>
<tr>
<td>$\pi^{**}$</td>
<td>Inflation target that minimizes the average loss function over the posterior distribution of $\theta$</td>
</tr>
</tbody>
</table>

B The distribution of optimal inflation targets

Figure B.1: Posterior Distribution of $\pi^*$ - EA - Benchmark

Note: Plain curve: PDF of $\pi^*$; Dashed vertical line: Average value of $\pi^*$ over posterior distribution; Dotted vertical line: Optimal inflation at the posterior mean of $\theta$; Dashed-dotted vertical line: Bayesian-theoretic optimal inflation
C Impact of a decline in the natural rate on the welfare criterion

Figures C.1a and C.1b below provide a more precise sense of how $\pi^*$ is modified following a decrease of $r^*$.

Figure C.1: $W(\pi, \mathbb{E}(\theta))$

(a) US

(b) EA
D Further illustration of the $(r^*, \pi^*)$ relation

D.1 When $\mu_z$ varies

Figure D.1: $(r^*, \pi^*)$ locus when $\mu_z$ varies

(a) US

(b) EA

Note: Blue: parameters set at the posterior mean; light blue: parameters set at the posterior median; Lighter blue: parameters set at the posterior mode. Memo: $r^* = \rho + \mu_z$. Range for $\mu_z$: 0.4% to 10% (annualized).

D.2 When $\rho$ varies

Figure D.2: $(r^*, \pi^*)$ locus when $\rho$ varies

(a) US

(b) EA

Note: Blue: parameters set at the posterior mean; light blue: parameters set at the posterior median; Lighter blue: parameters set at the posterior mode. Memo: $r^* = \rho + \mu_z$. Range for $\mu_z$: 0.4% to 10% (annualized).
E  Nominal and Real Interest Rates

Figure E.1: \((r^*, i^*)\) locus (at the posterior mean)

(a) US

(b) EA

Note: the blue dots correspond to the \((r^*, i^*)\) locus when \(r^*\) varies with \(\mu_z\); the red dots correspond to the \((r^*, i^*)\) locus when \(r^*\) varies with \(\rho\)

F  Distribution of \(\pi^*\) following a downward shift of the distribution of \(r^*\)

Figure F.1: Counterfactual - US

Note: The dashed vertical line indicates the mean value, i.e. \(E_\theta(\pi^*(\theta))\).
G Model Solution

G.1 Households

G.1.1 First Order Conditions

The associated lagrangian of program (1) under constraint (2) is

\[
\mathcal{L}_t = \mathbb{E}_t \sum_{s=0}^{\infty} \beta^s \left\{ e^{\xi_{t+s}} \log(C_{t+s} - \tilde{\eta}C_{t+s-1}) - \frac{\chi}{1+\nu} \int_0^1 e^{\xi_{t+s}} (N_{t+s}(h))^{1+\nu} dh \\
- \frac{\Lambda_{t+s}}{P_{t+s}} \left[ P_{t+s} C_{t+s} + Q_{t+s} B_{t+s} e^{-\tilde{\eta}q_{t+s}} + P_{t+s} \text{tax}_{t+s} - \int_0^1 W_{t+s}(h) N_{t+s}(h) dh - B_{t+s-1} - P_{t+s} \text{div}_{t+s} \right] \right\},
\]

The associated first order condition with respect to bonds is

\[
\frac{\partial \mathcal{L}_t}{\partial B_t} = 0 \Leftrightarrow \Lambda_t Q_t e^{-\tilde{\eta}q_t} = \beta \mathbb{E}_t \left\{ \frac{\Lambda_{t+1}}{\Pi_{t+1}} \right\},
\]

and the first-order condition with respect to consumption is

\[
\frac{\partial \mathcal{L}_t}{\partial C_t} = 0 \Leftrightarrow \frac{e^{\xi_{t}}} {C_t - \tilde{\eta}C_{t-1}} - \beta \tilde{\eta} \mathbb{E}_t \left\{ \frac{e^{\xi_{t+1}}}{C_{t+1} - \eta C_t} \right\} = \Lambda_t,
\]

where we defined

\[
\eta = \hat{\eta} e^{-\mu_z}.
\]
Let us define $i_t \equiv -\log(Q_t)$ and for any generic variable $X_t$

$$x_t \equiv \log(X_t), \quad \hat{x}_t \equiv x_t - x$$

where $x$ is the steady-state value of $x$. Using these definitions, log-linearizing equation (G.3) yields

$$\hat{g}_t + \beta \eta \mathbb{E}_t \{ \hat{\epsilon}_{t+1} \} - (1 + \beta \eta^2) \hat{\epsilon}_t + \eta \hat{\epsilon}_{t-1} - \eta(\zeta_{z,t} - \beta \mathbb{E}_t \{ \zeta_{z,t+1} \}) = \varphi^{-1} \hat{\lambda}_t \quad (G.5)$$

where we defined

$$\varphi^{-1} \equiv (1 - \beta \eta)(1 - \eta).$$

$$\hat{\epsilon}_t = (1 - \eta)(\zeta_{c,t} - \beta \eta \mathbb{E}_t \{ \zeta_{c,t+1} \}).$$

Similarly, log-linearizing equation (G.4) yields

$$\hat{\lambda}_t = i_t + \mathbb{E}_t \{ \hat{\lambda}_{t+1} - \hat{\pi}_{t+1} - \zeta_{z,t+1} \} + \zeta_{q,t}. \quad (G.6)$$

### G.2 Firms

Expressing the demand function in normalized terms yields

$$Y_{z,t}(f) = \left( \frac{P_t(f)}{P_t} \right)^{-\theta_p} Y_{z,t},$$

In the case of a firm not drawn to re-optimize, this equation specializes to (in log-linear terms)

$$\hat{y}_{t+s}(f) - \hat{y}_{t+s} = \theta_p(\hat{\pi}_{t+s} - \hat{\pi}_{t+s} - \hat{p}_t^* (f)). \quad (G.7)$$

### G.2.1 Cost Minimization

The real cost of producing $Y_t(f)$ units of good of $f$ is

$$W_t L_t(f) = W_t \left( \frac{Y_t(f)}{Z_t} \right)^{\phi} \quad (G.8)$$

The associated real marginal cost is thus

$$S_t(f) = \varphi^{\phi} \frac{W_t}{P_t Z_t} \left( \frac{Y_t(f)}{Z_t} \right)^{\phi^{-1}} \quad (G.9)$$

It is useful at this stage to restate the production function in log-linearized terms:

$$\hat{y}_{z,t}(f) = \frac{1}{\phi} \hat{n}_t(f) \quad (G.10)$$
G.2.2 Price Setting of Intermediate Goods: Optimization

Firm $f$ chooses $P^*_t(f)$ in order to maximize

$$
E_t \sum_{s=0}^{\infty} (\beta^{s} \alpha_p) \Lambda_{t+s} \left\{ \left( 1 + \tau_{p,t+s} \right) \frac{V_{t+1}^p P_t^*(f)}{P_{t+s}} Y_{t,s}^* - S \left( Y_{t,s}^* \right) \right\},
$$

subject to the demand function

$$
Y_{t,s}^* = \left( \frac{V_{t+1}^p P_t^*(f)}{P_{t+s}} \right)^{-\theta_p} Y_{t,s},
$$

and the cost schedule (G.8), where $\Lambda_t$ is the representative household’s marginal utility of wealth, and $E_t \{ \cdot \}$ is the expectation operator conditional on information available as of time $t$. That $\Lambda_t$ appears in the above maximization program reflects the fact that the representative household is the ultimate owner of firm $f$.

The associated first-order condition is

$$
E_t \sum_{s=0}^{\infty} (\beta^{s} \alpha_p) \Lambda_{t+s} \left\{ \left( \frac{V_{t+1}^p P_t^*(f)}{P_{t+s}} \right)^{1-\theta_p} Y_{t+s} - \frac{\mu_p}{1 + \tau_p} e^{s_{u,t+s}} W_{t+s} \phi \left( \left( \frac{V_{t+1}^p P_t^*(f)}{P_{t+s}} \right)^{-\theta_p} Y_{t+s} \right) \right\} = 0,
$$

where

$$
\mu_p = \frac{\theta_p}{\theta_p - 1}.
$$

This rewrites

$$
\left( \frac{P^*_t(f)}{P_t} \right)^{1+\theta_p(\phi-1)} = \frac{\mu_p}{1 + \tau_p} \frac{K_{p,t}}{F_{p,t}},
$$

where

$$
K_{p,t} = E_t \sum_{s=0}^{\infty} (\beta^{s} \alpha_p) \Lambda_{z,t+s} e^{s_{u,t+s}} W_{z,t+s} \phi \left( \left( \frac{V_{t+1}^p P_t^*(f)}{P_{t+s}} \right)^{-\theta_p} Y_{z,t+s} \right) \phi
$$

and

$$
F_{p,t} = E_t \sum_{s=0}^{\infty} (\beta^{s} \alpha_p) \Lambda_{z,t} \left( \frac{V_{t+1}^p P_t^*(f)}{P_{t+s}} \right)^{1-\theta_p} Y_{z,t+s},
$$

where $\Pi_{t+s} \equiv P_{t+s}/P_t$.

Notice that

$$
K_{p,t} = \phi \Lambda_{z,t} e^{s_{u,t}} W_{z,t} (Y_{z,t})^\phi + \beta \alpha_p E_t \left( \frac{[(\Pi)_{t+1}^{1-\gamma}]_{\nu} \gamma_p}{\Pi_{t+1}^{1-\gamma}} \right)^{-\phi \theta_p} K_{p,t+1},
$$

and

$$
F_{p,t} = \Lambda_{z,t} Y_{z,t} + \beta \alpha_p E_t \left( \frac{[(\Pi)_{t+1}^{1-\gamma}]_{\nu} \gamma_p}{\Pi_{t+1}^{1-\gamma}} \right)^{1-\theta_p} F_{p,t+1}.
$$

With a slight abuse of notation, we obtain the steady-state relation

$$
\left( \frac{P^*_t}{P} \right)^{1+\theta_p(\phi-1)} = \frac{\mu_p}{1 + \tau_p} \phi \frac{W_z^\phi}{P} Y_{z}^{\phi-1} \frac{1 - \beta \alpha_p (\Pi)_{t+1}^{1-\gamma \phi} (\theta_p - 1)}{1 - \beta \alpha_p (\Pi)_{t+1}^{1-\gamma \phi} (1-\gamma_p)}.
$$
Log-linearizing yields
\[ [1 + \theta_p(\phi - 1)](p_t^* - p_t) = \hat{k}_{p,t} - \hat{f}_{p,t} \]

\[ \hat{k}_{p,t} = (1 - \omega_{K,p})[\hat{\lambda}_{z,t} + \hat{\omega}_t + \phi\hat{g}_{z,t} + \hat{\zeta}_{u,t}] + \omega_{K,p}E_t\{\hat{k}_{p,t+1} + \phi\theta_p(\hat{n}_{t+1} - \iota_p\gamma_p\hat{n}_t)\}, \]

and
\[ \hat{f}_{p,t} = (1 - \omega_{F,p})(\hat{\lambda}_{z,t} + \hat{g}_{z,t}) + \omega_{F,p}E_t\{\hat{f}_{p,t+1} + (\theta_p - 1)(\hat{n}_{t+1} - \iota_p\gamma_p\hat{n}_t)\}, \]

where we defined the de-trended real wage
\[ \omega_t \equiv w_{z,t} - p_t \]

\[ \omega_{K,p} \equiv \beta\alpha_p(\Pi(1-\gamma_p)\phi_{\theta_p}) \]

and
\[ \omega_{F,p} \equiv \beta\alpha_p(\Pi(1-\gamma_p)(\theta_p-1)) \]

Finally, notice that
\[ p_t^{1-\theta_p} = \int_0^1 p_t(f)^{1-\theta_p} df \]

\[ = (1 - \alpha_p)(p_t^*)^{1-\theta_p} + \alpha_p \int_0^1 \left[ ((\Pi)^{1-\iota_p}(\Pi_{t-1})^\gamma_p P_t-1(f)) \right]^{1-\theta_p} df \]

Thus
\[ 1 = (1 - \alpha_p) \left( \frac{p_t^*}{P_t} \right)^{1-\theta_p} + \alpha_p \left[ \left( (\Pi)^{1-\iota_p}(\Pi_{t-1})^\gamma_p \right)^{1-\theta_p} \right] \]

The steady-state relation is
\[ \left( \frac{p_t^*}{P_t} \right)^{1-\theta_p} = \frac{1 - \alpha_p(\Pi(1-\gamma_p)(\theta_p-1))}{1 - \alpha_p} \]

Log-linearizing this yields
\[ \hat{\gamma}_t^* = \frac{\omega_{F,p}}{\beta - \omega_{F,p}}(\hat{n}_t - \iota_p\gamma_p\hat{n}_{t-1}). \]

G.3 Unions

G.3.1 Wage Setting

Union $h$ sets $W_t^*(h)$ so as to maximize
\[ \mathbb{E}_t \sum_{s=0}^{\infty} (\beta\alpha_w)^s \left\{ (1 + \tau_w) \frac{\Lambda_{t+s}}{P_{t+s}} e^{\gamma_z h^z} V_{t,s}^w W_t^*(h) N_{t,t+s}(h) - \frac{\chi}{1 + \nu} e^{\hat{\gamma}_z h^z} (N_{t,t+s}(h))^{1+\nu} \right\} , \]

where
\[ N_{t,t+s}(h) = \left( \frac{e^{\gamma_z h^z} V_{t,s}^w W_t^*(h)}{W_{t+s}} \right)^{-\theta_w} N_{t+s} \]
The associated first-order condition is

\[ \mathbb{E}_t \sum_{s=0}^{\infty} (\beta \alpha_w)^s \left\{ \Lambda_t \frac{W_{t+s}}{P_{t+s}} N_{t+s} \left( \frac{e^{\gamma_z \mu_z^W W_{t+s}^*} V_{t+s}^w}{\Pi_{t+s}} \right)^{1-\theta_w} - \frac{\mu_w}{1 + \tau_w \chi} e^{s \theta_w} \left( \frac{e^{\gamma_z \mu_z^W W_{t+s}^*} V_{t+s}^w}{\Pi_{t+s}} \right)^{-(1+v)\theta_w} \right\} N_{t+s}^1 = 0 \]

where \( \Pi_{t+s}^w = W_{t+s} / W_t \).

Rearranging yields

\[ \left( \frac{W_{t+s}^*}{W_t} \right)^{1+\theta_w} = \frac{\mu_w}{1 + \tau_w \chi} \frac{K_{w,t}}{F_{w,t}}. \]

where

\[ K_{w,t} = \mathbb{E}_t \sum_{s=0}^{\infty} (\beta \alpha_w)^s \left\{ \chi e^{s \theta_w} N_{t+s}^{1+v} + \beta \alpha_w \mathbb{E}_t \left( e^{\gamma_z \mu_z^W [(\Pi)_1^{1-\theta_w} (\Pi)_{t+s}^{1-\theta_w}] \gamma_w } \right)^{- (1+v) \theta_w} K_{w,t+1} \right\} \]

and

\[ F_{w,t} = \Lambda_t \frac{W_{t+s}}{P_{t+s}} N_{t+s} \left( \frac{e^{\gamma_z \mu_z^W V_{t+s}^w}}{\Pi_{t+s}} \right)^{1-\theta_w} \]

where \( \Pi_{t+s}^w \equiv W_{t+s} / W_t \).

Notice that

\[ K_{w,t} = e^{s \theta_w} N_{t}^{1+v} + \beta \alpha_w \mathbb{E}_t \left\{ \left( e^{\gamma_z \mu_z^W [(\Pi)_1^{1-\theta_w} (\Pi)_{t+s}^{1-\theta_w}] \gamma_w } \right)^{- (1+v) \theta_w} K_{w,t+1} \right\} \]

and

\[ F_{w,t} = \Lambda_t \frac{W_{t+s}}{P_{t+s}} N_{t+s} + \beta \alpha_w \mathbb{E}_t \left\{ \left( e^{\gamma_z \mu_z^W [(\Pi)_1^{1-\theta_w} (\Pi)_{t+s}^{1-\theta_w}] \gamma_w } \right)^{1-\theta_w} F_{w,t+1} \right\}. \]

The associated steady-state relations are

\[ \left( \frac{W_{t+s}^*}{W} \right)^{1+\theta_w} = \frac{\mu_w}{1 + \tau_w \chi} \frac{K_w}{F_w}. \]

\[ K_w = \frac{\chi N_{t+s}^{1+v}}{1 - \beta \alpha_w [e^{(1-\gamma_w)\mu_z^W (\Pi)_{t+s}^{1-\gamma_w}]}^{(1+v) \theta_w}. \]

\[ F_w = \frac{\Lambda_t^w \hat{H}}{1 - \beta \alpha_w [e^{(1-\gamma_w)\mu_z^W (\Pi)_{t+s}^{1-\gamma_w}]}^{\theta_w-1}. \]

Log-linearizing the above equations finally yields

\[ (1 + \theta_w) (w_{t+s}^* - w_t) = \hat{k}_{w,t} - f_{w,t}. \]

\[ \hat{k}_{w,t} = (1 - \omega_{K,w}) [(1 + v) \hat{n}_{t} + \zeta_{h,t}] + \omega_{K,w} \mathbb{E}_t \{ \hat{k}_{w,t+1} + (1 + v) \theta_w (\hat{n}_{w,t+1} - t_w \gamma_w \hat{n}_{t}) \} \]

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\[
\begin{align*}
\hat{f}_{w,t} &= (1 - \omega_{F,\omega})(\lambda_{z,t} + \omega_t + \hat{n}_t) + \omega_{F,\omega} E_t \{ \hat{f}_{w,t+1} + (\theta_w - 1)(\hat{\pi}_{w,t+1} - \iota_w \gamma_w \hat{\tau}_t) \},
\end{align*}
\]
where we defined
\[
\omega_{K,\omega} = \beta \alpha_{\omega} (\gamma z^\mu (\Pi) (1 - \gamma_w) (1 + \nu) \theta_w - 1)
\]
and
\[
\omega_{F,\omega} = \beta \alpha_{\omega} [e(1 - \gamma z) \mu z (\Pi) (1 - \gamma_w) (1 + \nu) \theta_w - 1]
\]
To complete this section, notice that
\[
1 = \left( 1 - \alpha_w \right) \left( \frac{W^*_t}{W_t} \right)^{1 - \theta_w} + \alpha_w \left( e^{\gamma z \mu z (\Pi) (1 - \gamma_w) (1 + \nu) \theta_w - 1} \right)
\]
and
\[
w^*_t - w_t = \frac{\omega_{F,\omega}}{\beta - \omega_{F,\omega}} (\hat{\pi}_{w,t} - \iota_w \gamma_w \hat{\tau}_{t-1}).
\]

### G.4 Market Clearing

The clearing on the labor market implies
\[
N_t = \left( \frac{Y_t}{Z_t} \right)^\phi \int_0^1 \left( \frac{P_t(f)}{P_t} \right)^{-\phi \theta_p} \, df.
\]
Let us define
\[
\Xi_{p,t} = \left( \int_0^1 \left( \frac{P_t(f)}{P_t} \right)^{-\phi \theta_p} \, df \right)^{-1/(\phi \theta_p)},
\]
so that
\[
N_t = (Y_t \Xi_{p,t}^{\theta_p})^\phi.
\]
Hence, expressed in log-linear terms, this equation reads
\[
\hat{n}_t = \phi (\hat{y}_{z,t} - \theta_p \hat{\xi}_{p,t}).
\]
Notice that
\[
\Xi_{p,t}^{-\phi \theta_p} = (1 - \alpha_p) \left[ P_t^{* -\phi \theta_p} \Pi_t \right] \left( [\Pi(1 - \iota_p (\Pi_{t-1})^{\iota_p} \gamma_p) - \phi \theta_p \Xi_{p,t-1}^{\phi \theta_p} \right]
\]
The associated steady-state relation is
\[
\Xi_{p,t}^{-\phi \theta_p} = \frac{(1 - \alpha_p)}{1 - \alpha_p (1 - \gamma_p (1 - \gamma_p) (1 - \gamma_p) \phi \theta_p)} \left( P_t^{* -\phi \theta_p} \right).
\]
Log-linearizing the price dispersion yields
\[
\hat{\xi}_{p,t} = (1 - \omega_{\Xi})(p_{t}^{*} - p_t) + \omega_{\Xi} [\hat{\xi}_{p,t-1} - (\hat{\pi}_t - \iota_p \gamma_p \hat{\tau}_{t-1})]
\]
where we defined
\[
\omega_{\Xi} = \alpha_p (1 - \gamma_p (1 - \gamma_p) \phi \theta_p).
\]
G.5 Natural Rate of Output

The natural rate of output is the level of production that would prevail in an economy without nominal rigidities, i.e. \( \alpha_p = \alpha_w = 0 \) and without cost-push shocks (i.e., \( \zeta_{u,t} = 0 \)). Under such circumstances, the dynamic system simplifies to

\[
\dot{\omega}_{z,t} + (\varphi - 1) \dot{y}_{z,t}^n = 0,
\]

\[
\nu \dot{n}_t + \zeta_{h,t} = \lambda_{z,t}^n + \dot{\omega}_{z,t},
\]

\[
\dot{n}_t = \phi \dot{y}_{z,t}^n,
\]

\[
\dot{g}_t + \beta \eta \mathbb{E}_t \{ \dot{y}_{z,t+1}^n \} - (1 + \beta \eta^2) \dot{y}_{z,t}^n + \eta \dot{y}_{z,t-1}^n - \eta (\zeta_{z,t} - \beta \mathbb{E}_t \{ \zeta_{z,t+1} \}) = \varphi^{-1} \lambda_{z,t}^n,
\]

where the superscript \( n \) stands for natural.

Combining these equations yields

\[
[\varphi (1 + \beta \eta^2) + \omega] \dot{y}_{z,t}^n - \varphi \beta \eta \mathbb{E}_t \{ \dot{y}_{z,t+1}^n \} - \varphi \eta \dot{y}_{z,t-1}^n = \varphi \dot{g}_t - \zeta_{h,t} - \varphi \eta \zeta_{z,t}^* + \omega \dot{y}_{z,t}^n - \varphi \beta \eta \mathbb{E}_t \{ \zeta_{z,t+1} \}
\]

where we defined

\[
\omega \equiv \nu \varphi + \varphi - 1,
\]

and

\[
\zeta_{z,t}^* = \zeta_{z,t} - \beta \mathbb{E}_t \{ \zeta_{z,t+1} \}
\]

G.6 Working Out the Steady State

The steady state is defined by the following set of equations

\[
1 - \beta \eta \frac{1}{(1 - \eta)C} = \Lambda_z,
\]

\[
e^{-i} = \beta e^{-\mu \cdot \Pi^{-1}},
\]

\[
\left( \frac{P^*}{P} \right)^{1 + \theta_p (\varphi - 1)} = \frac{\mu_p}{1 + \tau_p F_p} K_p
\]

\[
K_p = \frac{\phi \Lambda_z Y_p^\phi}{1 - \beta \alpha_p (\Pi)^{\theta_p (1 - \gamma_p)}},
\]
\[ F_p = \frac{\Lambda_z Y_z}{1 - \beta \alpha_p (\Pi)^{(1 - \gamma_p)(\theta_p - 1)}}. \]

\[ \left( \frac{P^*}{P} \right)^{1 - \theta_p} = \frac{1 - \alpha_p (\Pi)^{(1 - \gamma_p)(\theta_p - 1)}}{1 - \alpha_p}, \]

\[ \left( \frac{W^*}{W} \right)^{1 + \theta_w} = \frac{\mu_w}{1 + \tau_w} \frac{K_w}{F_w}. \]

\[ K_w = \frac{\chi^N (1 + v)}{1 - \beta \alpha_w [e^{(1 - \gamma_z)\mu_z (\Pi)^{1 - \gamma_w}}]^{(1 + v)\theta_w - 1}}, \]

\[ F_w = \frac{\Lambda_z W_z H}{1 - \beta \alpha_w [e^{(1 - \gamma_z)\mu_z (\Pi)(1 - \gamma_w)}]^{\theta_w - 1}}, \]

\[ \left( \frac{W^*}{W} \right)^{1 - \theta_w} = \frac{1 - \alpha_w [e^{(1 - \gamma_z)\mu_z (\Pi)(1 - \gamma_w)}]^{\theta_w - 1}}{1 - \alpha_w}, \]

\[ \Pi_w = \Pi e^{\mu_z}. \]

We can solve for \( i \) and \( \Pi_w \) using

\[ \Pi_w = \Pi e^{\mu_z}. \]

\[ 1 = \beta e^{-\mu_z} e^i \Pi^{-1}, \]

Standard manipulations yield

\[ 1 - \omega_{K,p} \left( \frac{\beta (1 - \alpha_p)}{\beta - \omega_{F,p}} \right)^{1 + \theta_p (\phi - 1)} = \frac{\mu_p}{1 + \tau_p} \frac{W_z Y_z^{\phi - 1}}{P}, \]

where we used

\[ \omega_{K,p} = \beta \alpha_p (\Pi)^{(1 - \gamma_p)(\theta_p)} \]

\[ \omega_{F,p} = \beta \alpha_p (\Pi)^{(1 - \gamma_p)(\theta_p - 1)} \]

Similar manipulations yield

\[ 1 - \omega_{K,w} \left( \frac{\beta (1 - \alpha_w)}{\beta - \omega_{F,w}} \right)^{1 + \theta_w v} = \frac{\mu_w}{1 + \tau_w} \frac{\chi^N v}{\Lambda_z W_z}, \]

where we used

\[ \omega_{K,w} = \beta \alpha_w [e^{(1 - \gamma_z)\mu_z (\Pi)(1 - \gamma_w)}]^{(1 + v)\theta_w} \]
Combining these conditions yields

\[
\frac{1 - \omega_{K,w}}{1 - \omega_{F,w}} \left( \frac{\beta(1 - \alpha_w)}{\beta - \omega_{F,p}} \right)^{\frac{1 + \theta_p}{\eta_p - 1}} = \frac{\mu_w}{1 + \tau_w} \frac{\mu_p}{1 + \tau_p} \frac{1 - \eta}{1 - \beta \eta} \phi \chi N^v Y_z^\phi
\]

Now, recall that

\[
(Y_z \Sigma_p^{-\theta})^\phi = N
\]

Then, using

\[
\Sigma_p^{-\phi} = \frac{1 - \alpha_p}{1 - \omega_z} \left( \frac{P^*}{\bar{P}} \right)^{-\phi},
\]

and

\[
\left( \frac{P^*}{\bar{P}} \right)^{-\phi} = \left( \frac{\beta(1 - \alpha_p)}{\beta - \omega_{F,p}} \right)^{-\phi},
\]

we end up with

\[
N^v Y_z^\phi = \left( \frac{1 - \alpha_p}{1 - \omega_z} \left( \frac{\beta(1 - \alpha_p)}{\beta - \omega_{F,p}} \right)^{-\phi} \right)^\nu Y_z^{(1+v)\phi},
\]

so that

\[
\Omega = \frac{\mu_w}{1 + \tau_w} \frac{\mu_p}{1 + \tau_p} \frac{1 - \eta}{1 - \beta \eta} \phi \chi Y_z^{(1+v)\phi},
\]

where

\[
\Omega = \frac{1 - \omega_{K,w}}{1 - \omega_{F,w}} \left( \frac{\beta(1 - \alpha_w)}{\beta - \omega_{F,p}} \right)^{\frac{1 + \theta_p}{\eta_p - 1}} \frac{1 - \omega_{K,p}}{1 - \omega_{F,p}} \left( \frac{\beta(1 - \alpha_p)}{\beta - \omega_{F,p}} \right)^{\frac{1 + \theta_p}{\eta_p - 1}} \left( \frac{1 - \omega_z}{1 - \alpha_p} \right)^\nu
\]

We defined the natural rate of output as the level of production that would prevail in an economy without nominal rigidities, i.e. \( \alpha_p = \alpha_w = 0 \). Under such circumstances, the steady-state value of the (normalized) natural rate of output \( y^n \) obeys

\[
1 = \frac{\mu_w}{1 + \tau_w} \frac{\mu_p}{1 + \tau_p} \frac{1 - \eta}{1 - \beta \eta} \phi \chi (Y_z^n)^{(1+v)}.
\]

It follows that the steady-state distortion due to sticky prices and wages (and less than perfect indexation) is

\[
\left( \frac{Y_z}{Y_z^n} \right)^{(1+v)} = \Omega.
\]

**H Welfare**

Let us define for any generic variable \( X_t \)

\[
\frac{X_t - X}{X} = \hat{x}_t + \frac{1}{2} \hat{\xi}_t^2 + \mathcal{O}(||\xi||^3)
\]
\[
\frac{X_t - X^n}{X^n} = \bar{x}_t + \frac{1}{2} \bar{x}_t^2 + O(||\bar{\zeta}||^3)
\]

Below, we repeatedly use the following two lemmas:

**Lemma 1.** Let \( g(\cdot) \) be a twice differentiable function and let \( X \) be a stationary random variable. Then
\[
E\{g(X)\} = g(E\{X\}) + \frac{1}{2} g''(E\{X\})V\{X\} + O(||X||^3).
\]

**Lemma 2.** Let \( g(\cdot) \) be a twice differentiable function and let \( x \) be a stationary random variable. Then
\[
V\{g(X)\} = [g'(E\{X\})]^2 V\{X\} + O(||X||^3).
\]

In the rest of this section, we take a second-order approximation of welfare, where we consider the inflation rate as an expansion parameter. It follows that we consider the welfare effects of non-zero trend inflation only up to second order.

### H.1 Second-Order Approximation of Utility

Consider first the utility derived from consumption. For the sake of notational simplicity, define
\[
U(C_{z,t} - \eta C_{z,t-1} e^{-\bar{\zeta}_{z,t}}) = \log(C_{z,t} - \eta C_{z,t-1} e^{-\bar{\zeta}_{z,t}})
\]
We thus obtain
\[
e^{\bar{\zeta}_{z,t}} U(C_{z,t} - \eta C_{z,t-1} e^{-\bar{\zeta}_{z,t}}) = \frac{1}{1 - \eta} \left[ \left( \frac{C_{z,t} - C^n_z}{C^n_z} \right) - \eta \left( \frac{C_{z,t-1} - C^n_z}{C^n_z} \right) \right]
- \frac{1}{2} \frac{1}{(1 - \eta)} \left( \frac{C_{z,t} - C^n_z}{C^n_z} \right)^2 + \frac{\eta}{(1 - \eta)} \left( \frac{C_{z,t} - C^n_z}{C^n_z} \right) \left( \frac{C_{z,t-1} - C^n_z}{C^n_z} \right) - \frac{1}{2} \frac{\eta^2}{(1 - \eta)} \left( \frac{C_{z,t-1} - C^n_z}{C^n_z} \right)^2
+ \bar{\zeta}_{c,t} \left( \frac{C_{z,t} - C^n_z}{C^n_z} \right) - \eta \bar{\zeta}_{c,t} \left( \frac{C_{z,t-1} - C^n_z}{C^n_z} \right) - \frac{\eta}{(1 - \eta)} \bar{\zeta}_{z,t} \left( \frac{C_{z,t} - C^n_z}{C^n_z} \right) + \frac{\eta}{(1 - \eta)} \bar{\zeta}_{z,t} \left( \frac{C_{z,t-1} - C^n_z}{C^n_z} \right)
+ \text{t.i.p} + O(||\bar{\zeta}||^3),
\]
where t.i.p stands for terms independent of policy.

Then, using
\[
\frac{C_{z,t} - C^n_z}{C^n_z} = \bar{c}_{z,t} + \frac{1}{2} \bar{c}_{z,t}^2 + O(||\bar{\zeta}||^3)
\]
we obtain
\[
e^{\bar{\zeta}_{z,t}} U(C_{z,t} - \eta C_{z,t-1} e^{-\bar{\zeta}_{z,t}}) = \frac{1}{1 - \eta} \left[ \bar{c}_{z,t} - \eta \bar{c}_{z,t-1} + \frac{1}{2} (\bar{c}_{z,t} - \eta \bar{c}_{z,t-1}) \right]
- \frac{1}{2} \frac{1}{1 - \eta} \bar{c}_{z,t}^2 + \frac{\eta}{1 - \eta} \bar{c}_{z,t} \bar{c}_{z,t-1} - \frac{1}{2} \frac{\eta^2}{1 - \eta} \bar{c}_{z,t}^2 - \frac{1}{2} \frac{\eta^2}{1 - \eta} \bar{c}_{z,t-1}^2
+ \bar{\zeta}_{c,t} (\bar{c}_{z,t} - \eta \bar{c}_{z,t-1}) - \frac{\eta}{1 - \eta} \bar{\zeta}_{z,t} (\bar{c}_{z,t} - \bar{c}_{z,t-1})
+ \text{t.i.p} + O(||\bar{\zeta}||^3),
\]
Using

\[ \varphi^{-1} = (1 - \beta \eta)(1 - \eta) \]

we obtain

\[
e^{\tilde{c}_{z,t}} U(C_{z,t} - \eta C_{z,t-1} e^{-\tilde{c}_{z,t}}) = \frac{1}{1 - \eta} \left[ \tilde{y}_{z,t} - \eta \tilde{y}_{z,t-1} + \frac{1}{2} ( \tilde{y}_{z,t}^2 - \eta \tilde{y}_{z,t-1}^2 ) \right] - \frac{1}{2} (1 - \beta \eta) \varphi \tilde{y}_{z,t}^2 + \eta (1 - \beta \eta) \varphi \tilde{y}_{z,t} \tilde{y}_{z,t-1} - \frac{1}{2} \eta^2 (1 - \beta \eta) \varphi \tilde{y}_{z,t}^2 + \tilde{\zeta}_{c,t} (\tilde{y}_{z,t} - \eta \tilde{y}_{z,t-1}) - \eta (1 - \beta \eta) \varphi \tilde{\zeta}_{z,t} (\tilde{y}_{z,t} - \tilde{y}_{z,t-1}) \right] + \text{t.i.p} + O(||\zeta||^3),
\]

where we imposed the equilibrium condition on the goods market.

Similarly, taking a second-order approximation of labor disutility in the neighborhood of the natural steady-state \( N^n \) yields

\[
\frac{X}{1 + v} e^{\tilde{c}_{h,t}} (N_t(h))^{1+v} = \chi(N^n)^{1+v} \left( \frac{N_t(h) - N^n}{N^n} \right) + \frac{1}{2} \chi^v (N^n)^{1+v} \left( \frac{N_t(h) - N^n}{N^n} \right)^2 \\
+ \chi(N^n)^{1+v} \left( \frac{N_t(h) - N^n}{N^n} \right) \tilde{\zeta}_{h,t} + \text{t.i.p} + O(||\zeta||^3).
\]

Now, using

\[
\frac{N_t(h) - N^n}{N^n} = \tilde{n}_t(h) + \frac{1}{2} \bar{n}_t(h)^2 + O(||\zeta||^3)
\]

we get

\[
\frac{X}{1 + v} e^{\tilde{c}_{h,t}} (N_t(h))^{1+v} = \chi(N^n)^{1+v} \left[ \tilde{n}_t(h) + \frac{1}{2} (1 + v) \bar{n}_t(h)^2 + \tilde{n}_t(h) \tilde{\zeta}_{h,t} \right] + \text{t.i.p} + O(||\zeta||^3).
\]

Integrating over the set of labor types, one gets

\[
\int_0^1 \frac{X}{1 + v} e^{\tilde{c}_{h,t}} (N_t(h))^{1+v} dh = \chi(N^n)^{1+v} \left[ \mathbb{E}_h \{ \tilde{n}_t(h) \} + \frac{1}{2} (1 + v) \mathbb{E}_h \{ \bar{n}_t(h)^2 \} + \mathbb{E}_h \{ \bar{n}_t(h) \} \tilde{\zeta}_{h,t} \right] + \text{t.i.p} + O(||\zeta||^3).
\]

Now, since

\[
\mathbb{V}_h \{ \tilde{n}_t(h) \} = \mathbb{E}_h \{ \tilde{n}_t(h)^2 \} - \mathbb{E}_h \{ \bar{n}_t(h) \}^2,
\]

the above relation rewrites

\[
\int_0^1 \frac{X}{1 + v} e^{\tilde{c}_{h,t}} (N_t(h))^{1+v} dh = \chi(N^n)^{1+v} \left[ \mathbb{E}_h \{ \tilde{n}_t(h) \} + \frac{1}{2} (1 + v) (\mathbb{V}_h \{ \tilde{n}_t(h) \} + \mathbb{E}_h \{ \bar{n}_t(h) \}^2) \\
+ \mathbb{E}_h \{ \bar{n}_t(h) \} \tilde{\zeta}_{h,t} \right] + \text{t.i.p} + O(||\zeta||^3).
\]

We need to express \( \mathbb{E}_h \{ \tilde{n}_t(h) \} \) and \( \mathbb{V}_h \{ \tilde{n}_t(h) \} \) in terms of the aggregate variables. To this end, we first establish a series of results, on which we draw later on.
H.2 Aggregate Labor and Aggregate Output

Notice that

$$\frac{\theta_w - 1}{\theta_w} \tilde{n}_t = \log \left( \int_0^1 \left( \frac{N_t(h)}{N^n} \right)^{(\theta_w - 1)/\theta_w} \, dh \right).$$

Then, applying lemma 1, one obtains

$$\tilde{n}_t = \mathbb{E}_h \{ \tilde{n}_t(h) \} + \frac{1}{2} \frac{\theta_w}{\theta_w - 1} \mathbb{E}_h \left\{ \left( \frac{N_t(h)}{N^n} \right)^{\frac{\theta_w - 1}{\theta_w}} \right\}^2 \mathbb{V}_h \left\{ \left( \frac{N_t(h)}{N^n} \right)^{\frac{\theta_w - 1}{\theta_w}} \right\} + \mathcal{O}(||\zeta||^3).$$

Then, notice that

$$\mathbb{V}_h \left\{ \left( \frac{N_t(h)}{N^n} \right)^{\frac{\theta_w - 1}{\theta_w}} \right\} = \mathbb{V}_h \left\{ \exp \left[ (1 - \theta_w^{-1}) \log \left( \frac{N_t(h)}{N^n} \right) \right] \right\}$$

so that, by applying lemma 2, one obtains

$$\mathbb{V}_h \left\{ \left( \frac{N_t(h)}{N^n} \right)^{\frac{\theta_w - 1}{\theta_w}} \right\} = (1 - \theta_w^{-1})^2 \exp \left[ (1 - \theta_w^{-1}) \mathbb{E}_h \{ \tilde{n}_t(h) \} \right]^2 \mathbb{V}_h \{ \tilde{n}_t(h) \} + \mathcal{O}(||\zeta||^3).$$

Similarly

$$\mathbb{E}_h \left\{ \left( \frac{N_t(h)}{N^n} \right)^{\frac{\theta_w - 1}{\theta_w}} \right\} = \mathbb{E}_h \left\{ \exp \left[ (1 - \theta_w^{-1}) \tilde{n}_t(h) \right] \right\}$$

so that, by applying lemma 1 once more, one obtains

$$\mathbb{E}_h \left\{ \left( \frac{N_t(h)}{N^n} \right)^{\frac{\theta_w - 1}{\theta_w}} \right\} = \exp \left[ (1 - \theta_w^{-1}) \mathbb{E}_h \{ \tilde{n}_t(h) \} \right] \left( 1 + \frac{1}{2} (1 - \theta_w^{-1})^2 \mathbb{V}_h \{ \tilde{n}_t(h) \} \right) + \mathcal{O}(||\zeta||^3).$$

Then combining the previous results

$$\tilde{n}_t = \mathbb{E}_h \{ \tilde{n}_t(h) \} + \frac{1}{2} \frac{1}{1 - \theta_w^{-1}} \frac{(1 - \theta_w^{-1})^2 \mathbb{V}_h \{ \tilde{n}_t(h) \}}{\left( 1 + \frac{1}{2} (1 - \theta_w^{-1})^2 \mathbb{V}_h \{ \tilde{n}_t(h) \} \right)^2} + \mathcal{O}(||\zeta||^3).$$

It is convenient to define

$$\Delta_{h,t} \equiv \mathbb{V}_h \{ \tilde{n}_t(h) \}$$

so that finally

$$\tilde{n}_t = \mathbb{E}_h \{ \tilde{n}_t(h) \} + Q_{0,h} + \frac{1 - \theta_w^{-1}}{2} Q_{1,h} (\Delta_{h,t} - \Delta_n) + \mathcal{O}(||\zeta||^3).$$

where we defined

$$Q_{0,h} = \frac{\left[ 1 - \frac{1}{2} (1 - \theta_w^{-1})^2 \Delta_n \right]^2}{\left[ 1 + \frac{1}{2} (1 - \theta_w^{-1})^2 \Delta_n \right]^2}$$

and

$$Q_{1,h} = \frac{1 - \frac{1}{2} (1 - \theta_w^{-1})^2 \Delta_n}{\left[ 1 + \frac{1}{2} (1 - \theta_w^{-1})^2 \Delta_n \right]^3}$$

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Applying the same logic on output and defining

$$\Delta y, t \equiv V_f \{ \tilde{y}_t(f) \}$$

one gets

$$\tilde{y}_{z,t} = E_f \{ \tilde{y}_{z,t}(f) \} + Q_{0,y} + \frac{1 - \theta_p^{-1}}{2}Q_{1,y} \Delta y, t - \Delta y + O(||\xi||^3).$$

where we defined

$$Q_{0,y} = \frac{1 - \theta_p^{-1}}{2} \frac{\Delta y}{\left[ 1 + \frac{1}{2}(1 - \theta_p^{-1})^2 \Delta y \right]^2}$$

and

$$Q_{1,y} = \frac{1 - \frac{1}{2}(1 - \theta_p^{-1})^2 \Delta y}{\left[ 1 + \frac{1}{2}(1 - \theta_p^{-1})^2 \Delta y \right]^3}$$

Then recall from (??) and from the equilibrium on the market for aggregate labor that

$$N_t = \int_0^1 L_t(f) \, df = \int_0^1 Y_{z,t}(f) \, df$$

which implies

$$\frac{N_t}{N^n} = \int_0^1 \left( \frac{Y_{z,t}(f)}{Y^n} \right) \phi \, df$$

where we used $N^n = (Y^n_z)^\phi$.

This relation rewrites

$$\hat{n}_t = \log \left( \int_0^1 \left( \frac{Y_{z,t}(f)}{Y^n_z} \right)^\phi \, df \right)$$

This expression is of the form

$$\hat{n}_t = \log \left( E_f \left\{ \left( \frac{Y_{z,t}(f)}{Y^n_z} \right)^\phi \right\} \right) .$$

Using lemmas 1 and 2, we obtain the following three approximations

$$\hat{n}_t = E_f \{ \phi(\tilde{y}_{z,t}(f) - z_t) \} + \frac{1}{2} \frac{\nabla_f \left\{ \left( \frac{Y_{z,t}(f)}{Y^n_z} \right)^\phi \right\} }{E_f \left\{ \left( \frac{Y_{z,t}(f)}{Y^n_z} \right)^\phi \right\} } + O(||\xi||^3),$$

$$\nabla_f \left\{ \left( \frac{Y_{z,t}(f)}{Y^n_z} \right)^\phi \right\} = \phi^2 \left[ \exp \left[ \phi E_f \{ \tilde{y}_{z,t}(f) \} \right] \right] \nabla_f \{ \tilde{y}_{z,t}(f) \} + O(||\xi||^3),$$

$$E_f \left\{ \left( \frac{Y_{z,t}(f)}{Y^n_z} \right)^\phi \right\} = \exp \left[ \phi E_f \{ \tilde{y}_{z,t}(f) \} \right] \left( 1 + \frac{1}{2} \phi^2 \nabla_f \{ \tilde{y}_{z,t}(f) \} \right) + O(||\xi||^3).$$

Combining these expressions as before yields

$$\hat{n}_t = \phi E_f \{ \tilde{y}_{z,t}(f) \} + \frac{1}{2} \phi^2 \frac{\nabla_f \{ \tilde{y}_{z,t}(f) \} }{(1 + \frac{1}{2} \phi^2 \nabla_f \{ \tilde{y}_{z,t}(f) \})^2} + O(||\xi||^3).$$
We finally obtain
\[ \bar{n}_t = \phi \mathbb{E}_f \{ \bar{z}_{l,t}(f) \} + P_{0,y} + \frac{1}{2} \phi^2 P_{1,y}(\Delta y,y) + \mathcal{O}(||\tilde{\zeta}||^3), \]
where we used
\[ \frac{\nabla_f \{ \bar{z}_{l,t}(f) \}}{(1 + \frac{1}{2} \phi^2 \nabla_f \{ \bar{z}_{l,t}(f) \})^2} = \frac{\Delta y}{(1 + \frac{1}{2} \phi^2 \Delta y)^2} + \frac{1 - \frac{1}{2} \phi^2 \Delta y}{(1 + \frac{1}{2} \phi^2 \Delta y)^3} (\Delta y,y) + \mathcal{O}(||\tilde{\zeta}||^3) \]
and defined
\[ P_{0,y} = \frac{\frac{1}{2} \phi^2 \Delta y}{(1 + \frac{1}{2} \phi^2 \Delta y)^2} \]
and
\[ P_{1,y} = \frac{1 - \frac{1}{2} \phi^2 \Delta y}{(1 + \frac{1}{2} \phi^2 \Delta y)^3} \]

H.3 Aggregate Price and Wage Levels

The aggregate price index is
\[ P_t^{1-\theta_p} = \left( \int_0^1 P_t(f)^{1-\theta_p} df \right) \]
and the aggregate wage index is
\[ W_t^{1-\theta_w} = \left( \int_0^1 W_t(h)^{1-\theta_w} dh \right). \]

From lemma 1 and the definitions of \( P_t \) and \( W_t \), we obtain
\[ p_t = \mathbb{E}_f \{ p_t(f) \} + \frac{1}{2} \frac{1}{1 - \theta_p} \mathbb{V}_f \{ P_t(f)^{1-\theta_p} \} + \mathcal{O}(||\tilde{\zeta}||^3), \]
and
\[ w_t = \mathbb{E}_h \{ w_t(h) \} + \frac{1}{2} \frac{1}{1 - \theta_w} \mathbb{V}_h \{ W_t(h)^{1-\theta_w} \} + \mathcal{O}(||\tilde{\zeta}||^3). \]

Then, from lemma 2, we obtain
\[ \mathbb{V}_f \{ P_t(f)^{1-\theta_p} \} = \mathbb{V}_f \{ \exp[(1 - \theta_p) P_t(f)] \} = (1 - \theta_p)^2 \exp[(1 - \theta_p) p_t]^2 \Delta p_t \Delta y + \mathcal{O}(||\tilde{\zeta}||^3), \]
and
\[ \mathbb{V}_h \{ W_t(h)^{1-\theta_w} \} = \mathbb{V}_h \{ \exp[(1 - \theta_w) W_t(h)] \} = (1 - \theta_w)^2 \exp[(1 - \theta_w) w_t]^2 \Delta w_t \Delta y + \mathcal{O}(||\tilde{\zeta}||^3), \]
where we defined
\[ \bar{p}_t = \mathbb{E}_f \{ p_t(f) \}, \quad \bar{w}_t = \mathbb{E}_h \{ w_t(h) \}, \]
\[ \Delta_{p,t} = \mathbb{V}_f \{ p_t(f) \}, \quad \Delta_{w,t} = \mathbb{V}_h \{ w_t(h) \}. \]

Applying lemma 1 once again, we obtain
\[
\mathbb{E}_f \{ P_t(f)^{1-\theta} \} = \mathbb{E}_f \{ \exp[(1 - \theta_p) p_t(f)] \} = \exp[(1 - \theta_p) \bar{p}_t] \left( 1 + \frac{1}{2} (1 - \theta_p)^2 \Delta_{p,t} \right)
\]

and
\[
\mathbb{E}_h \{ W_t(h)^{1-\theta_w} \} = \mathbb{E}_h \{ \exp[(1 - \theta_w) w_t(h)] \} = \exp[(1 - \theta_w) \bar{w}_t] \left( 1 + \frac{1}{2} (1 - \theta_w)^2 \Delta_{w,t} \right)
\]

Combining these relations, we obtain
\[
p_t = \bar{p}_t + \frac{1}{2} \frac{(1 - \theta_p) \Delta_{p,t}}{\left[ 1 + \frac{1}{2} (1 - \theta_p)^2 \Delta_{p,t} \right]^2} + \mathcal{O}(||\zeta||^3),
\]

and
\[
w_t = \bar{w}_t + \frac{1}{2} \frac{(1 - \theta_w) \Delta_{w,t}}{\left[ 1 + \frac{1}{2} (1 - \theta_w)^2 \Delta_{w,t} \right]^2} + \mathcal{O}(||\zeta||^3).
\]

Thus
\[
p_t = \bar{p}_t + Q_{0,p} + \frac{1 - \theta_p}{2} Q_{1,p} (\Delta_{p,t} - \Delta_p) + \mathcal{O}(||\zeta||^3),
\]

and
\[
w_t = \bar{w}_t + Q_{0,w} + \frac{1 - \theta_w}{2} Q_{1,w} (\Delta_{w,t} - \Delta_w) + \mathcal{O}(||\zeta||^3).
\]

where we defined
\[
Q_{0,p} = \frac{1 - \theta_p}{2} \frac{\Delta_p}{\left[ 1 + \frac{1}{2} (1 - \theta_p)^2 \Delta_p \right]^2}, \quad Q_{0,w} = \frac{1 - \theta_w}{2} \frac{\Delta_w}{\left[ 1 + \frac{1}{2} (1 - \theta_w)^2 \Delta_w \right]^2}
\]

and
\[
Q_{1,p} = \frac{1 - \frac{1}{2} (1 - \theta_p)^2 \Delta_p}{\left[ 1 + \frac{1}{2} (1 - \theta_p)^2 \Delta_p \right]^3}, \quad Q_{1,w} = \frac{1 - \frac{1}{2} (1 - \theta_w)^2 \Delta_w}{\left[ 1 + \frac{1}{2} (1 - \theta_w)^2 \Delta_w \right]^3}
\]

Remark that the constant terms in the second-order approximation of the log-price index can be rewritten as
\[
Q_{0,p} - \frac{1 - \theta_p}{2} Q_{1,p} \Delta_p = \frac{1}{2} \frac{(1 - \theta_p)^3 \Delta_p^2}{\left[ 1 + \frac{1}{2} (1 - \theta_p)^2 \Delta_p \right]^3}.
\]

Finally, using the demand functions, one obtains
\[
\tilde{y}_{z,t}(f) = -\theta_p [p_t(f) - p_t] + \tilde{y}_{z,t},
\]
\[ \tilde{n}_t(h) = -\theta_w[w_t(h) - w_t] + \tilde{n}_t, \]
from which we deduce that
\[ \Delta_{y,t} = \theta_p^2 \Delta_{p,t} \]
and
\[ \Delta_{h,t} = \theta_w^2 \Delta_{w,t}. \]

**H.4 Price and Wage Dispersions**

We now derive the law of motion of price dispersion. Notice that
\[ \Delta_{p,t} = \Psi_f \{ p_t(f) - \bar{p}_{t-1} \} \]

Immediate manipulations of the definition of the cross-sectional mean of log-prices yield
\[ \bar{p}_t - \bar{p}_{t-1} = \alpha_p \gamma_p [(1 - t_p) \pi + t_p \pi_{t-1}] + (1 - \alpha_p) [p_{t-1}^* - \bar{p}_{t-1}]. \] \hfill (H.1)

Then, the classic variance formula yields
\[ \Delta_{p,t} = \mathbb{E}_f \{ [p_t(f) - \bar{p}_{t-1}]^2 \} - \mathbb{E}_f \{ p_t(f) - \bar{p}_{t-1} \}^2 \]

Using this, we obtain
\[ \Delta_{p,t} = \alpha_p \mathbb{E}_f \{ [p_{t-1}(f) - \bar{p}_{t-1} + \gamma_p \tilde{\pi}_{t-1}]^2 \} + (1 - \alpha_p) [p_{t-1}^* - \bar{p}_{t-1}]^2 - [\bar{p}_t - \bar{p}_{t-1}]^2 \]

where we defined
\[ \tilde{\pi}_t \equiv \pi + t_p (\pi_t - \pi) \]

Notice that
\[ (1 - \alpha_p) [p_{t-1}^* - \bar{p}_{t-1}]^2 - [\bar{p}_t - \bar{\bar{p}}_{t-1}]^2 \]
\[ = (1 - \alpha_p) \left[ \frac{1}{1 - \alpha_p} (\bar{p}_t - \bar{p}_{t-1}) - \frac{\alpha_p}{1 - \alpha_p} \gamma_p \tilde{\pi}_t \right]^2 - [\bar{p}_t - \bar{p}_{t-1}]^2 \]
\[ = \frac{\alpha_p}{1 - \alpha_p} [\bar{p}_t - \bar{p}_{t-1} - \gamma_p \tilde{\pi}_t]^2 - \alpha_p [\gamma_p \tilde{\pi}_t]^2 \]

Using this in the above equation yields
\[ \Delta_{p,t} = \alpha_p \mathbb{E}_f \{ [p_{t-1}(f) - \bar{p}_{t-1} + \gamma_p \tilde{\pi}_{t-1}]^2 \} - \alpha_p [\gamma_p \tilde{\pi}_{t-1}]^2 + \frac{\alpha_p}{1 - \alpha_p} [\bar{p}_t - \bar{p}_{t-1} - \gamma_p \tilde{\pi}_t]^2 \]

Now, notice also that
\[ \alpha_p \mathbb{E}_f \{ [p_{t-1}(f) - \bar{p}_{t-1}]^2 \} = \alpha_p \mathbb{E}_f \{ [p_{t-1}(f) - \bar{p}_{t-1} + \gamma_p \tilde{\pi}_{t-1}]^2 \} - \alpha_p [\gamma_p \tilde{\pi}_t]^2 \]
It then follows that
\[
\Delta_{p,t} = \alpha_p \mathbb{E}_f \{ [p_{t-1}(f) - \bar{p}_{t-1}]^2 \} + \frac{\alpha_p}{1 - \alpha_p} [\bar{p}_t - \bar{p}_{t-1} - \gamma_p \bar{p}_t]^2
\]

Hence
\[
\Delta_{p,t} = \alpha_p \Delta_{p,t-1} + \frac{\alpha_p}{1 - \alpha_p} [\bar{p}_t - \bar{p}_{t-1} - \gamma_p \bar{p}_t]^2
\]

which, in turn, implies
\[
\Delta_{p,t} = \alpha_p \Delta_{p,t-1} + \frac{\alpha_p}{1 - \alpha_p} [\bar{p}_t - \bar{p}_{t-1} - \gamma_p (\pi + \iota_p (\bar{\pi}_{t-1} - \pi))]^2.
\]

Using
\[
p_t = p_t + Q_{0,p} + \frac{1 - \theta_p}{2} Q_{1,p} (\Delta_{p,t} - \Delta_p) + \mathcal{O}(||\xi||^3),
\]
we obtain
\[
\bar{p}_t - \bar{p}_{t-1} = \pi_t - \frac{1 - \theta_p}{2} Q_{1,p} (\Delta_{p,t} - \Delta_{p,t-1}) + \mathcal{O}(||\xi||^3).
\]

Hence
\[
\Delta_{p,t} = \alpha_p \Delta_{p,t-1} + \frac{\alpha_p}{1 - \alpha_p} \left[ \pi_t - \frac{1 - \theta_p}{2} Q_{1,p} (\Delta_{p,t} - \Delta_{p,t-1}) - \gamma_p (\pi + \iota_p (\bar{\pi}_{t-1} - \pi)) \right]^2 + \mathcal{O}(||\xi||^3).
\]

The steady-state value of \(\Delta_p\) is thus
\[
\Delta_p = \frac{(1 - \gamma_p)^2 \alpha_p}{(1 - \alpha_p)^2} \pi^2
\]

We obtain finally
\[
\Delta_{p,t} = \alpha_p \Delta_{p,t-1} + \frac{\alpha_p}{1 - \alpha_p} \left[ (1 - \gamma_p) \pi + \bar{\pi}_t - \gamma_p \iota_p \bar{\pi}_{t-1} - \frac{1 - \theta_p}{2} Q_{1,p} (\Delta_{p,t} - \Delta_{p,t-1}) \right]^2 + \mathcal{O}(||\xi||^3).
\]

Unless \(\pi\) is itself treated as an expansion variable in the above approximation, we cannot claim that \(\Delta_{p,t}\) is second-order.

We now derive the law of motion of wage dispersion. Following similar steps as for price dispersion, notice that
\[
\Delta_{w,t} = \mathbb{V}_h \{ w_t(h) - \bar{w}_{t-1} \}
\]

Immediate manipulations of the definition of the cross-sectional mean of log-wages yield
\[
\bar{w}_t - \bar{w}_{t-1} = \alpha_w (\gamma_z \mu_z + \gamma_w [(1 - \iota_w) \pi + \iota_w \bar{\pi}_{t-1}]) + (1 - \alpha_w) [w_t^* - \bar{w}_{t-1}]. \tag{H.2}
\]

Then, the classic variance formula yields
\[
\Delta_{w,t} = \mathbb{E}_h \{ [w_t(h) - \bar{w}_{t-1}]^2 \} - \mathbb{E}_h \{ w_t(h) - \bar{w}_{t-1} \}^2
\]
Using this, we obtain
\[
\Delta_{w,t} = \alpha_w \mathbb{E}_h \{ [w_{t-1}(h) - \bar{w}_{t-1} + \gamma_z \mu_z + \gamma_w \bar{\pi}_{t-1}]^2 \} + (1 - \alpha_w) [w_t^* - \bar{w}_{t-1}]^2 - [\bar{w}_t - \bar{w}_{t-1}]^2 
\]
where, this time, we defined
\[
\bar{\pi}_t = \pi + \iota_w(\pi_t - \pi)
\]
Notice that
\[
w_t^* - \bar{w}_{t-1} = \frac{1}{1 - \alpha_w} (\bar{w}_t - \bar{w}_{t-1}) - \frac{\alpha_w}{1 - \alpha_w} [\gamma_z \mu_z + \gamma_w \bar{\pi}_t]
\]
so that
\[
(1 - \alpha_w) [w_t^* - \bar{w}_{t-1}]^2 - [\bar{w}_t - \bar{w}_{t-1}]^2
\]
\[
= (1 - \alpha_w) \left[ \frac{1}{1 - \alpha_w} (\bar{w}_t - \bar{w}_{t-1}) - \frac{\alpha_w}{1 - \alpha_w} [\gamma_z \mu_z + \gamma_w \bar{\pi}_t] \right]^2 - [\bar{w}_t - \bar{w}_{t-1}]^2
\]
\[
= \frac{\alpha_w}{1 - \alpha_w} [\bar{w}_t - \bar{w}_{t-1} - [\gamma_z \mu_z + \gamma_w \bar{\pi}_t]]^2 - \alpha_w [\gamma_z \mu_z + \gamma_w \bar{\pi}_t]^2
\]
Using this in the above equation yields
\[
\Delta_{w,t} = \alpha_w \mathbb{E}_h \{ [w_{t-1}(h) - \bar{w}_{t-1} + \gamma_z \mu_z + \gamma_w \bar{\pi}_t]^2 \} 
\]
\[
- \alpha_w [\gamma_z \log(1 + \bar{\pi}_t)]^2 + \frac{\alpha_w}{1 - \alpha_w} [\bar{w}_t - \bar{w}_{t-1} - [\gamma_z \mu_z + \gamma_w \bar{\pi}_t]]^2
\]
Now, notice also that
\[
\alpha_w \mathbb{E}_h \{ [w_{t-1}(h) - \bar{w}_{t-1}]^2 \} = \alpha_w \mathbb{E}_h \{ [w_{t-1}(h) - \bar{w}_{t-1} + \gamma_z \mu_z + \gamma_w \bar{\pi}_t]^2 \} - \alpha_w [\gamma_z \mu_z + \gamma_w \bar{\pi}_t]^2
\]
It then follows that
\[
\Delta_{w,t} = \alpha_w \mathbb{E}_h \{ [w_{t-1}(h) - \bar{w}_{t-1}]^2 \} + \frac{\alpha_w}{1 - \alpha_w} \left[ \bar{w}_t - \bar{w}_{t-1} - [\gamma_z \mu_z + \gamma_w \bar{\pi}_t] \right]^2
\]
Hence
\[
\Delta_{w,t} = \alpha_w \Delta_{w,t-1} + \frac{\alpha_w}{1 - \alpha_w} \left[ \bar{w}_t - \bar{w}_{t-1} - [\gamma_z \mu_z + \gamma_w \bar{\pi}_t] \right]^2
\]
which, in turn, implies
\[
\Delta_{w,t} = \alpha_w \Delta_{w,t-1} + \frac{\alpha_w}{1 - \alpha_w} [\bar{w}_t - \bar{w}_{t-1} - \gamma_z \mu_z - \gamma_w (\pi + \iota_w(\pi_t - \pi))]^2.
\]
Using
\[
\bar{w}_t = \bar{w}_t + Q_{0,w} + \frac{1 - \theta_w}{2} Q_{1,w}(\Delta_{w,t} - \Delta_{w}) + \mathcal{O}(||\zeta||^3),
\]
we obtain
\[
\bar{w}_t - \bar{w}_{t-1} = \pi_{w,t} - \frac{1 - \theta_w}{2} Q_{1,w}(\Delta_{w,t} - \Delta_{w,t-1}) + \mathcal{O}(||\zeta||^3).
\]
Hence

\[
\Delta_{w,t} = \Delta_{w,t-1} + \frac{\alpha_w}{1 - \alpha_w} \left[ \pi_{w,t} - \frac{1 - \theta_w}{2} Q_{1,w} (\Delta_{w,t} - \Delta_{w,t-1}) \right.
\]

\[- \gamma_z \mu_z - \gamma_w (\pi + \lambda_w (\pi_{t-1} - \pi)) \right] + O(||\zeta||^3).
\]

The steady-state value of \(\Delta_w\) is thus

\[
\Delta_w = \frac{\alpha_w}{(1 - \alpha_w)^2} [(1 - \gamma_z) \mu_z + (1 - \gamma_w) \pi]^2
\]

We obtain finally

\[
\Delta_{w,t} = \Delta_{w,t-1} + \frac{\alpha_w}{1 - \alpha_w} \left[ (1 - \gamma_z) \mu_z + (1 - \gamma_w) \pi + \mu_{w,t} - \gamma_w \mu_{t-1} \right.
\]

\[- \frac{1 - \theta_w}{2} Q_{1,w} (\Delta_{w,t} - \Delta_{w,t-1}) \right] + O(||\zeta||^3).
\]

Here, treating \(\pi\) as an expansion parameter will not suffice to ensure that wage dispersion \(\Delta_{w,t}\) is second order because an additional constant term linked to average productivity shows up in the above equations. Henceforth, we will assume that \(\mu_z\) is an expansion parameter.

Then, if we treat \(\pi\) as an expansion variable, precisely because the steady-state value of \(\Delta_p\) is of second-order, many of the expressions previously derived considerably simplify. In particular, we now obtain

\[
p_t = P_t + \frac{1 - \theta_p}{2} \Delta_{p,t} + O(||\zeta, \pi||^3),
\]

\[
w_t = W_t + \frac{1 - \theta_w}{2} \Delta_{w,t} + O(||\zeta, \pi||^3),.
\]

Now, because \(\Delta_y\) and \(\Delta_h\) are proportional to \(\Delta_p\) and \(\Delta_w\), respectively, and because \(\Delta_p\) and \(\Delta_w\) are both proportional to \(\pi^2\), we also obtain

\[
\bar{n}_t = \mathbb{E}_h \{ \bar{n}_t (h) \} + \frac{1 - \theta_w^{-1}}{2} \Delta_{h,t} + O(||\zeta, \pi||^3),
\]

\[
\bar{n}_t = \mathbb{E}_f \{ \bar{y}_z,t (f) \} - z_t + \frac{1}{2} \phi^2 \Delta_{y,t} + O(||\zeta, \pi||^3),
\]

\[
\bar{y}_t = \mathbb{E}_f \{ \bar{y}_t (f) \} + \frac{1 - \theta_p^{-1}}{2} \Delta_{y,t} + O(||\zeta, \pi||^3),.
\]

Thus, for sufficiently small inflation rates, we obtain formulas resembling those derived in Woodford (2003).
Finally, price and wage dispersions rewrite

\[ \Delta_{p,t} = \alpha_p \Delta_{p,t-1} + \frac{\alpha_p}{1-\alpha_p} \left[ (1 - \gamma_p) \pi + \pi_t - \gamma_p \pi_t^t \right] + O(||\zeta, \pi||^3), \]

\[ \Delta_{w,t} = \alpha_w \Delta_{w,t-1} + \frac{\alpha_w}{1-\alpha_w} \left[ (1 - \gamma_w) \mu + (1 - \gamma_w) \pi + \pi_{w,t} - \gamma_w \pi_{w,t}^t \right] + O(||\zeta, \pi||^3), \]

### H.5 Combining the Results

Combining the previous results, we obtain

\[
\int_0^1 \frac{\chi}{1+\nu} e^{e^{\xi_{l,t}} (N_t(h))^{1+v}} dh = \chi(N^v) \left[ \hat{n}_t + \frac{1}{2}(1+\nu)\hat{n}_t^2 + \hat{n}_t \tilde{\zeta}_{h,t} \right] + \frac{1}{2}(1+\nu\theta_w)\theta_w \Delta_{w,t} + t.i.p + O(||\zeta, \pi||^3),
\]

In turn, we have

\[ \hat{n}_t = \phi \hat{y}_t + \frac{1}{2} \phi [(\phi - 1) \theta_p + 1] \theta_p \Delta_{p,t} + O(||\zeta, \pi||^3), \]

so that

\[
\int_0^1 \frac{\chi}{1+\nu} e^{e^{\xi_{l,t}} (N_t(h))^{1+v}} dh = \phi \chi(N^v) \left[ (\hat{y}_t - z_t) + \frac{1}{2}(1+\nu)\phi \hat{y}_t^2 + \hat{y}_t \tilde{\zeta}_{h,t} \right] + \frac{1}{2} [(\phi - 1) \theta_p + 1] \theta_p \Delta_{p,t} + \frac{1}{2} (1+\nu\theta_w)\phi^{-1} \theta_w \Delta_{w,t} + t.i.p + O(||\zeta, \pi||^3),
\]

Then, using

\[ (1 - \Phi) \frac{1 - \beta \eta}{1 - \eta} = \phi \chi(N^v), \]

where we defined

\[ 1 - \Phi \equiv \frac{1 + \tau_w}{\mu_w} \frac{1 + \tau_p}{\mu_p} \]

we obtain

\[
\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ \int_0^1 \frac{\chi}{1+\nu} e^{e^{\xi_{l,t}} (N_t(h))^{1+v}} dh \right\} =
\]

\[ (1 - \Phi) \frac{1 - \beta \eta}{1 - \eta} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left[ \hat{y}_t + \frac{1}{2}(1+\nu)\phi \hat{y}_t^2 + \hat{y}_t \tilde{\zeta}_{h,t} \right] + \frac{1}{2} [(\phi - 1) \theta_p + 1] \theta_p \Delta_{p,t} + \frac{1}{2} (1+\nu\theta_w)\phi^{-1} \theta_w \Delta_{w,t} + t.i.p + O(||\zeta, \pi||^3). \]
Assuming the distortions are themselves negligible, this simplifies further to

$$
\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ \int_0^1 \frac{X}{1 + \nu} e^{\xi_{s,t}} (N_t(h))^{1+v} \, dh \right\} =
$$

$$
\frac{1 - \beta \eta}{1 - \eta} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ (1 - \Phi)\tilde{y}_t + \frac{1}{2} (1 + v)\phi \tilde{y}_t^2 + \tilde{y}_t \tilde{\zeta}_{h,t} 
\right. 
+ \left. \frac{1}{2} [(\phi - 1)\theta_p + 1]\theta_p \Delta_{p,t} + \frac{1}{2} (1 + v\theta_w)\phi^{-1}\theta_w \Delta_{w,t} \right\} + t.i.p + O(||\zeta||^3),
$$

We now deal with the first term in the utility function. To that end, notice that

$$
\sum_{t=0}^{\infty} \beta^t a_{t-1} = a_{-1} + \beta \sum_{t=0}^{\infty} \beta^t a_{t-1} = a_{-1} + \beta \sum_{t=0}^{\infty} \beta^t a_t.
$$

Using this trick, we obtain

$$
\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t e^{\xi_{s,t}} \log(C_{z,t} - \eta C_{z,t-1} e^{-\xi_{z,t}}) = \frac{1 - \beta \eta}{1 - \eta} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ \tilde{g}_{z,t} - \frac{1}{2} [\varphi (1 + \beta \eta^2) - 1] \tilde{y}_t^2 + \eta \varphi \tilde{y}_t \tilde{y}_{z,t-1}
\right. 
+ \varphi \hat{\theta}_t \tilde{g}_{z,t} - \eta \varphi \tilde{\zeta}_{z,t} \tilde{y}_{z,t} + t.i.p + O(||\zeta||^3),
$$

where we defined

$$\varphi^{-1} = (1 - \beta \eta)(1 - \eta),$$

$$\hat{\theta}_t = (1 - \eta)(\tilde{\zeta}_{c,t} - \beta \eta \mathbb{E}_t \{\tilde{\zeta}_{c,t+1}\}),$$

so that

$$(1 - \beta \eta) \varphi \hat{\theta}_t \equiv (\tilde{\zeta}_{c,t} - \beta \eta \mathbb{E}_t \{\tilde{\zeta}_{c,t+1}\}).$$

and

$$\tilde{\zeta}^*_{z,t} = \tilde{\zeta}_{z,t} - \beta \mathbb{E}_t \{\tilde{\zeta}_{z,t+1}\}.$$

Combining terms, we obtain

$$U_0 = \frac{1 - \beta \eta}{1 - \eta} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ \Phi \tilde{g}_{z,t} - \frac{1}{2} [\varphi (1 + \beta \eta^2) + \omega] \tilde{g}^2_{z,t} + \eta \varphi \tilde{y}_t \tilde{y}_{z,t-1}
\right. 
+ \left. (\varphi \hat{\theta}_t - \tilde{\zeta}_{h,t} - \varphi \eta \tilde{\zeta}^*_{z,t}) \tilde{y}_{z,t}
\right. 
- \left. \frac{1}{2} [(\phi - 1)\theta_p + 1]\theta_p \Delta_{p,t} - \frac{1}{2} (1 + v\theta_w)\phi^{-1}\theta_w \Delta_{w,t} \right\} + t.i.p + O(||\zeta||^3),
$$

where, as defined earlier

$$\omega = (1 + v)\phi - 1.$$
Now, recall that

\[ [\varphi (1 + \beta \eta^2) + \omega] y_{z,t}^n - \varphi \beta \eta E_i \{ y_{z,t+1}^n \} - \varphi \eta y_{z,t-1}^n = \varphi \hat{g}_t - \zeta_{h,t} - \varphi \eta \hat{g}^*_t \]

Using this above yields

\[
U_0 = \frac{1 - \beta \eta}{1 - \eta} E_0 \sum_{t=0}^{\infty} \beta^t \left[ \Phi \hat{y}_{z,t} - \frac{1}{2} [\varphi (1 + \beta \eta^2) + \omega] \hat{y}_t^2 + \eta \varphi \hat{y}_{z,t} \hat{y}_{z,t-1} \right]
\]

\[
+ [\varphi (1 + \beta \eta^2) + \omega] y_{z,t}^n \hat{y}_{z,t} - \varphi \beta \eta y_{z,t+1}^n \hat{y}_{z,t} - \varphi \eta y_{z,t-1}^n \hat{y}_{z,t}
\]

\[
- \frac{1}{2} [(\varphi - 1) \theta_p + 1] \theta_p \Delta_{p,t} - \frac{1}{2} (1 + v \theta_w) \Phi^{-1} \theta_w \Delta_{w,t} \right] + \text{t.i.p} + \mathcal{O}(|\zeta|^3),
\]

\[
U_0 = \frac{1 - \beta \eta}{1 - \eta} E_0 \sum_{t=0}^{\infty} \beta^t \left[ \Phi \hat{y}_t - \frac{1}{2} [\varphi (1 + \beta \eta^2) + \omega] \hat{y}_t^2 + \eta \varphi \hat{y}_t \hat{y}_{t-1} \right]
\]

\[
+ [\omega + \varphi (1 + \beta \eta^2)] \hat{y}_t \hat{y}_t - \varphi \beta \eta \hat{y}_{t+1}^n \hat{y}_t - \varphi \eta \hat{y}_{t-1}^n \hat{y}_t
\]

\[
- \frac{1}{2} [(\varphi - 1) \theta_p + 1] \theta_p \Delta_{p,t} - \frac{1}{2} (1 + v \theta_w) \Phi^{-1} \theta_w \Delta_{w,t} \right] + \text{t.i.p} + \mathcal{O}(|\zeta, \pi|^3)
\]

To simplify this expression, we seek constant terms \( \delta_0, \delta \) and \( x^* \) such that

\[
E_0 \sum_{t=0}^{\infty} \beta^t \left\{ - \frac{1}{2} \delta_0 [(\hat{y}_t - \hat{y}^n_t) - \delta (\hat{y}_{t-1} - \hat{y}^n_{t-1}) - x^*]^2 \right\}
\]

\[
= E_0 \sum_{t=0}^{\infty} \beta^t \left[ \Phi \hat{y}_t - \frac{1}{2} [\varphi (1 + \beta \eta^2) + \omega] \hat{y}_t^2 + \eta \varphi \hat{y}_t \hat{y}_{t-1} \right]
\]

\[
+ [\omega + \varphi (1 + \beta \eta^2)] \hat{y}_t \hat{y}_t - \varphi \beta \eta \hat{y}_{t+1}^n \hat{y}_t - \varphi \eta \hat{y}_{t-1}^n \hat{y}_t \right] + \text{t.i.p}
\]

Developing yields

\[
- \frac{\delta_0}{2} \left[ (\hat{y}_t - \hat{y}^n_t) - \delta (\hat{y}_{t-1} - \hat{y}^n_{t-1}) - x^* \right]^2
\]

\[
= - \frac{1}{2} \delta_0 \hat{y}_t^2 + \delta_0 \hat{y}_t \hat{y}^n_t + \delta_0 \delta \hat{y}_t \hat{y}_{t-1} - \delta_0 \delta \hat{y}_t \hat{y}^n_{t-1} - \delta_0 \delta \hat{y}_{t-1}^n \hat{y}_t
\]

\[
- \frac{1}{2} \delta_0 \delta^2 \hat{y}_{t-1}^2 + \delta_0 \delta^2 \hat{y}_{t-1} \hat{y}^n_{t-1} + \delta_0 (\hat{y}_t - \hat{y}_{t-1}) x^* + \text{t.i.p}
\]

Thus

\[
E_0 \sum_{t=0}^{\infty} \beta^t \left\{ - \frac{\delta_0}{2} [(\hat{y}_t - \hat{y}^n_t) - \delta (\hat{y}_{t-1} - \hat{y}^n_{t-1}) - x^*]^2 \right\}
\]

\[
= E_0 \sum_{t=0}^{\infty} \beta^t \left\{ \delta_0 (1 - \beta \delta)x^* \hat{y}_t - \frac{1}{2} \delta_0 (1 + \beta \delta^2) \hat{y}_t^2 + \delta_0 \delta \hat{y}_t \hat{y}_{t-1}
\]

\[
+ \delta_0 (1 + \beta \delta^2) \hat{y}_t \hat{y}^n_t - \delta_0 \delta \hat{y}_t \hat{y}^n_{t-1} - \delta_0 \delta \hat{y}_{t-1}^n \hat{y}_t + \text{t.i.p}
\]
Identifying term by term, we obtain
\[ \delta_0(1 - \beta \delta)x^* = \Phi \]

\[ \delta_0(1 + \beta \delta^2) = [\omega + \varphi(1 + \beta \eta^2)] \]

\[ \delta_0 \delta = \eta \varphi \]

Combining these relations, we obtain
\[ \eta \delta^2 - \frac{\omega + \varphi)(1 + \beta \eta^2)}{\beta \varphi} \delta + \eta \beta^{-1} = 0, \]

or equivalently
\[ \Pi(\kappa) = \beta^{-1} \kappa^2 - \chi \kappa + \eta^2 = 0, \]

where
\[ \kappa = \frac{\eta}{\delta}, \]

\[ \chi = \frac{\omega + \varphi(1 + \beta \eta^2)}{\beta \varphi} > 0. \]

Notice that
\[ \Pi(0) = \eta^2 > 0, \]

\[ \Pi(1) = -\frac{\omega}{\beta \varphi} < 0 \]

so that the two roots of \( \Pi(\kappa) = 0 \) obey
\[ 0 < \kappa_1 < 1 < \kappa_2. \]

In the sequel, we focus on the larger root and define
\[ \kappa = \kappa_2 = \frac{\beta}{2} \left( \chi + \sqrt{\chi^2 - 4 \eta^2 \beta^{-1}} \right) > 1. \]

Since \( \delta = \eta / \kappa \), we have
\[ 0 \leq \delta \leq \eta < 1. \]

Thus, given the obtained value for \( \kappa \), we can deduce \( \delta \) from which we can compute \( \delta_0 \).

We thus obtain
\[ U_0 = -\frac{1 - \beta \eta}{1 - \eta} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ \frac{\delta_0}{2} [(\tilde{y}_t - \hat{y}^H_t) - \delta(\tilde{y}_{t-1} - \hat{y}^H_{t-1}) - x^*]^2 \right. \]

\[ + \frac{1}{2} [(\phi - 1) \theta_p + 1] \theta_p \Delta_{p,t} + \frac{1}{2} (1 + v \theta_w) \phi^{-1} \theta_w \Delta_{w,t} \left\} + t.i.p + \mathcal{O}(|\zeta, \pi|^3), \]
The last step consists in expressing price and wage dispersions in terms of squared price and wage inflations.

Recall that

\[ \Delta_p, t = \alpha^p \Delta_p, t-1 + \frac{\alpha^p}{1 - \alpha^p} \left[ (1 - \gamma_p) \pi + \hat{\pi}_t - \gamma_p \hat{\pi}_t \hat{\pi}_{t-1} \right]^2 + \mathcal{O}(|\zeta, \pi|), \]

Iterating backward on this formula yields

\[ \Delta_p, t = \alpha^p \sum_{s=0}^{t} \alpha^p \left[ (1 - \gamma_p) \pi + \hat{\pi}_t - \gamma_p \hat{\pi}_t \hat{\pi}_{t-1} \right]^2 + \text{i.p} + \mathcal{O}(|\zeta, \pi|), \]

It follows that

\[ \sum_{t=0}^{\infty} \beta^t \Delta_{p,t} = \frac{\alpha^p}{(1 - \alpha^p)(1 - \beta \alpha^p)} \sum_{t=0}^{\infty} \beta^t \left[ (1 - \gamma_p) \pi + \hat{\pi}_t - \gamma_p \hat{\pi}_t \hat{\pi}_{t-1} \right]^2 + \text{i.p} + \mathcal{O}(|\zeta, \pi|), \]

and by the same line of reasoning

\[ \sum_{t=0}^{\infty} \beta^t \Delta_{w,t} = \frac{\alpha^w}{(1 - \alpha^w)(1 - \beta \alpha^w)} \sum_{t=0}^{\infty} \beta^t \left[ (1 - \gamma_z) \mu_z + (1 - \gamma_w) \pi + \hat{\pi}_t - \gamma_w \hat{\pi}_t \hat{\pi}_{t-1} \right]^2 + \text{i.p} + \mathcal{O}(|\zeta, \pi|), \]

Thus, defining

\[ \lambda_y \equiv \delta_0 \]

\[ \lambda_p \equiv \frac{\alpha^p \theta_p [(\phi - 1) \theta_p + 1]}{(1 - \alpha^p)(1 - \beta \alpha^p)} \]

\[ \lambda_w \equiv \frac{\alpha^w \phi^{-1} \theta_w (1 + v \theta_w)}{(1 - \alpha^w)(1 - \beta \alpha^w)} \]

The second order approximations to welfare rewrites

\[ U_0 = -\frac{1}{2} \frac{1 - \beta \eta}{1 - \eta} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left\{ \lambda_y [x_t - \delta x_{t-1} + (1 - \delta) \bar{x} - \bar{x}^*] \right\}^2 \]

\[ + \lambda_p [(1 - \gamma_p) \pi + \hat{\pi}_t - \gamma_p \hat{\pi}_t \hat{\pi}_{t-1}]^2 \]

\[ + \lambda_w [(1 - \gamma_z) \mu_z + (1 - \gamma_w) \pi + \hat{\pi}_t - \gamma_w \hat{\pi}_t \hat{\pi}_{t-1}]^2 + \text{i.p} + \mathcal{O}(|\zeta, \pi|), \]

where we defined

\[ x_t \equiv \hat{y}_t - \hat{y}_t^n \]

\[ \bar{x} \equiv \log \left( \frac{Y_z}{Y_w^n} \right). \]
References


