The Zero Lower Bound, the Dual Mandate, and Unconventional Dynamics

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Abstract

This article examines monetary policy when it is constrained by the zero lower bound (ZLB) on the nominal interest rate. Our analysis uses a nonlinear New Keynesian model with technology and discount factor shocks. Specifically, we investigate why technology shocks may have unconventional effects at the ZLB, what factors affect the likelihood of hitting the ZLB, and the implications of alternative monetary policy rules. We initially focus on a New Keynesian model without capital (Model 1) and then study that model with capital (Model 2). The advantage of including capital is that it introduces another mechanism for intertemporal substitution that strengthens the expectational effects of the ZLB. Four main findings emerge: (1) In Model 1, the choice of output target in the Taylor rule may reverse the effects of technology shocks when the ZLB binds; (2) When the central bank targets steady-state output in Model 2, a positive technology shock at the ZLB leads to more pronounced unconventional dynamics than in Model 1; (3) The presence of capital changes the qualitative effects of demand shocks and alters the impact of a monetary policy rule that emphasizes output stability; and (4) In Model 1, the constrained linear solution is a decent approximation of the nonlinear solution, but meaningful differences exist between the solutions in Model 2.

Keywords: Monetary Policy; Zero Lower Bound; Nonlinear Solution Method; Capital

JEL Classifications: E31; E42; E58; E61
1 Introduction

In the aftermath of the 2008 financial crisis, aggregate demand fell sharply. The Fed responded by lowering its policy rate to its zero lower bound (ZLB) by the end of the year. Six years after the crisis began, the Fed’s target interest rate remains near zero and the economy is below potential.

Figure 1 shows the U.S. and Japanese interbank lending rates and employment-to-population percentages from 1990-2014. The U.S. policy rate (solid line) has varied between 8.3% and 0% since 1990 and has been held below 25 basis points since the end of 2008. During that time period, policymakers shifted their focus from inflation to the real economy, since the inflation rate has been at or below the Fed’s inflation target. The Bank of Japan sharply lowered its policy rate in 1991 (dashed line), reaching 50 basis points in 1995. Since then it has remained between 0 and 50 basis points, while the employment-to-population percentage has fallen steadily from 62% to about 57.5%. The Japanese economy slightly rebounded in the mid-2000s, but after the financial crisis, the policy rate was cut and the employment-to-population percentage fell even further.

Figure 1: U.S. and Japanese interbank lending rates (left panel) and employment-to-population percentages (right panel). Sources: Federal Reserve, Bank of Japan, U.S. Bureau of Labor Statistics, and Statistics Bureau of Japan.

Over the last two decades, the Japanese economy has endured anemic growth in real GDP and slight deflation. Their experience has generated a significant amount of research on the effects of the Bank of Japan’s zero interest rate policy [e.g., Braun and Waki (2006); Eggertsson and Woodford (2003); Hoshi and Kashyap (2000); Ito and Mishkin (2006); Krugman (1998); Posen (1998)]. Many arguments for avoiding the ZLB are motivated in part by the recent Japanese experience.

This article examines the consequences of the ZLB constraint on the nominal interest rate. Our analysis uses a nonlinear New Keynesian model with technology and discount factor shocks that allows for the ZLB to occasionally bind. Discount factor shocks are a proxy for changes in demand that occurred during the Great Recession, while technology shocks account for changes in supply. When either shock pushes the nominal rate toward zero, households increasingly anticipate a ZLB event, which affects current economic outcomes through expectations. We refer to that anticipation as the “expectational effects” of hitting the ZLB. There are similar expectational effects of leaving the ZLB. Our solution method captures both of those effects. Within this framework, we investigate why technology shocks may have unconventional effects at the ZLB, what factors affect the likelihood of hitting the ZLB, and the tradeoffs a central bank faces under a dual mandate.
We initially focus on a New Keynesian model without capital and then study that model with capital to draw comparisons. In the model without capital, positive technology shocks may have unconventional effects at the ZLB, depending on which measure of output is targeted in the monetary policy rule. When the central bank targets steady-state output, positive technology shocks can cause output to decline when the ZLB binds. Those unconventional dynamics, however, nearly disappear when the central bank targets potential output, which is the level of output in our model with flexible prices. In that case, only large technology shocks reduce output when the ZLB binds. We show the differences between the two output targets since both are used in the literature.

We focus on the specification in which the central bank targets steady-state output, but it is optimal in our model to target potential output. The Fed’s January 2012 long-term policy statement emphasizes its dual mandate—stable prices and an economy operating at potential. Given that potential output is unobservable, policymakers tend to target an empirical measure of potential output that has the smooth characteristics of steady-state output [Basu and Fernald (2009)]. Moreover, Orphanides (2003a,b) and Orphanides and van Norden (2002) show a variety of estimates of potential output require substantial revisions as more data become available, which indicates potential output is not measured accurately in real time. For those reasons, we analyze the theoretical implications of targeting steady-state output and compare them to a potential output target.

Most of the ZLB literature uses models without capital. Capital, however, provides households with another margin to smooth consumption, which strengthens the expectational effects of the ZLB. Arbitrage implies the real interest rate equals the expected future real rental rate of capital. The decline in demand when the ZLB binds leads to a sharp reduction in the rental rate of capital. Therefore, households place increasing weight on the possibility of a lower future rental rate as the policy rate approaches zero, which causes sharper declines in the real interest rate before the ZLB binds. We also include capital adjustment costs to dampen investment volatility. That feature makes investment less attractive as a consumption smoothing mechanism, which causes a greater reduction in consumption and a larger increase in the real interest rate at the ZLB. When the central bank targets steady-state output, a positive technology shock at the ZLB produces more pronounced unconventional dynamics in our model with capital than in the model without capital.

We also evaluate how alternative monetary policy rules affect the likelihood of hitting the ZLB and the efficacy of stabilization policy. A policy rule based on a dual mandate is more likely to cause ZLB events when the central bank targets steady-state output in our model without capital. The opposite result occurs when the central bank targets potential output. When technology is constant, an aggressive response by the central bank to steady-state output decreases the frequency of ZLB events in our model without capital but increases the frequency in our model with capital.

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1Wieland (2014) uses structural VAR evidence to argue that these unconventional dynamics did not occur following the 2011 earthquake/tsunami in Japan or the recent oil supply shocks. Braun and Waki (2006) show technology shocks generate unconventional dynamics at the ZLB in a log-linearized model with capital where the central bank targets steady-state output. Using a nonlinear model with capital and a monetary policy rule that does not respond to output, Braun and Körber (2011) show that these unconventional dynamics may disappear if the expected duration at the ZLB is short enough. We find the monetary response to output also changes the qualitative effects of technology shocks.


3Several papers solve for the optimal monetary policy in a model with a ZLB constraint [Coenen et al. (2004); Eggertsson and Woodford (2003); Jung et al. (2005); Nakov (2008); Werning (2011)]. For example, Adam and Billi (2006) find that it is optimal to reduce the nominal interest rate more aggressively in response to adverse shocks.
Therefore, the frequency of ZLB events depends on (1) the measure of the output target; (2) the strength of the response to the output gap; and (3) the sources of exogenous shocks in the model.

Any analysis of the ZLB is complicated by the kink that it imposes on the monetary policy rule. The literature has used a variety of techniques to address this problem. Many papers separate the problem into pre- and post-ZLB periods [e.g., Braun and Körber (2011); Braun and Waki (2006); Christiano et al. (2011); Eggertsson and Woodford (2003); Ercgeg and Lindé (2014); Gertler and Karadi (2011)]. With that approach, a specific sequence of shocks pushes the nominal interest rate to zero. Each period, some positive probability exists that the nominal interest rate will exit the ZLB. Once that happens, the nominal interest rate can never fall back to zero. Those simplifying assumptions are made for computational tractability. The drawback is that if a shock causes the ZLB to bind in one period, the same shock will not cause the ZLB to bind in any future period.

Most studies of the ZLB linearize all of their equations with the exception of the monetary policy rule around their non-stochastic steady states. Such a procedure, however, can generate approximation errors. Braun et al. (2012) and Fernández-Villaverde et al. (2012) provide examples of the mistakes resulting from linearized models without capital evaluated at the ZLB. Braun et al. (2012) also argue that linearized models often lead to incorrect inferences about existence and uniqueness of the equilibrium and the local dynamics of the model. Our findings indicate the constrained linear model is a good approximation of the nonlinear model without capital, but the errors are much larger in a model with capital.\footnote{Braun and Waki (2010) show that the approximation error in a perfect-foresight version of a linear model with capital where monetary policy does not respond to output overstates the government spending multiplier.} In other words, the simulated moments and model predictions are different in the linearized model with capital than in the nonlinear model.

Our paper avoids the problems associated with linearization by obtaining the nonlinear solution to standard New Keynesian models that include an occasionally binding ZLB constraint on the nominal interest rate.\footnote{Several recent papers study the ZLB using nonlinear solution methods. Fernández-Villaverde et al. (2012) calculate the probabilities of ZLB events. Wolman (2005) shows the real effects of the ZLB depend on the policy rule and nominal rigidities. Gust et al. (2013) estimate the extent to which the ZLB constrained the central bank. Aruoba and Schorfheide (2013) and Mertens and Ravn (2014) show how the ZLB affects fiscal multipliers and Basu and Bundick (2012) and Nakata (2012) show the ZLB magnifies the effect of uncertainty on aggregate demand.} Rather than focus on specific sequences of shocks, we calculate the solution for all combinations of discount factor and technology shocks and then provide a thorough explanation of how dynamics change across the state space. Our nonlinear solution method emphasizes accuracy to capture important expectational effects of going to and returning from the ZLB.

The paper proceeds as follows. Section 2 outlines our models with and without capital. Section 3 describes the calibration and solution method, and sections 4 through 6 present the results. These sections report the model solutions across all technology and discount factor shocks, the dynamics at the ZLB, and the likelihood of hitting the ZLB. We also explain how the monetary policy rule impacts those results and provide a comparison between the New Keynesian models with and without capital. Lastly, we present new evidence that the solutions to the constrained linear and nonlinear models are significantly different in the model with capital. Section 7 concludes.

## 2 Economic Models

This section presents two New Keynesian models with Rotemberg (1982) price adjustment costs. Both models assume stochastic processes for the discount factor and technology, but they differ in their treatment of capital. Model 1 does not include capital while Model 2 does.
2.1 Model 1: Baseline

A representative household chooses \( \{c_t, n_t, b_t\}_{t=0}^{\infty} \) to maximize expected lifetime utility given by \( E_0 \sum_{t=0}^{\infty} \beta_t \log c_t - \chi n_t^{1+\eta}/(1+\eta) \), where \( 1/\eta \) is the Frisch elasticity of labor supply, \( c_t \) is consumption, \( n_t \) is labor hours, \( b_t \) is the real value of a 1-period nominal bond, \( E_0 \) is an expectation operator conditional on information available in period 0, \( \beta_0 \equiv 1 \), and \( \beta_t = \prod_{j=1}^{t} \beta_j \) for \( t > 0 \). \( \beta \) is a time-varying subjective discount factor that evolves according to

\[
\beta_t = \beta_t^{(1)} \beta_t^{(2)} \exp(\varepsilon_t),
\]

where \( \beta_t^{(1)} \) is the steady-state discount factor, \( 0 \leq \rho_\beta < 1 \), and \( \varepsilon_t \sim N(0, \sigma_\varepsilon^2) \). Those choices are constrained by \( c_t + b_t = w_t n_t + r_{t-1} b_{t-1}/\pi_t + d_t \), where \( \pi_t = p_t/p_{t-1} \) is the gross inflation rate, \( w_t \) is the real wage rate, \( r_t \) is the gross nominal interest rate set by the central bank, and \( d_t \) are profits from intermediate firms. The optimality conditions to the household’s problem imply

\[
w_t = \chi n_t^\eta c_t, \quad 1 = r_t E_t [\beta_{t+1} (c_t/c_{t+1})/\pi_{t+1}],
\]

(2)

The production sector consists of monopolistically competitive intermediate goods firms who produce a continuum of differentiated inputs and a representative final goods firm. Each firm \( f \in [0,1] \) in the intermediate goods sector produces a differentiated good, \( y_t(f) \), with identical technologies given by \( y_t(f) = z_t n_t(f) \), where \( n_t(f) \) is the level of employment used by firm \( f \). \( z_t \) represents the level of technology, which is common across firms and follows

\[
z_t = \bar{z}(z_{t-1}/\bar{z})^\rho z \exp(\nu_t),
\]

(3)

where \( \bar{z} \) is steady-state technology, \( 0 \leq \rho_z < 1 \), and \( \nu_t \sim N(0, \sigma^2_\nu) \). Each intermediate firm chooses its labor supply to minimize its operating costs, \( w_t n_t(f) \), subject to its production function. The final goods firm purchases \( y_t(f) \) units from each intermediate goods firm to produce the final good, \( y_t \equiv \int_0^1 y_t(f)(\theta-1)/\theta df \) according to a Dixit and Stiglitz (1977) aggregator, where \( \theta > 1 \) measures the elasticity of substitution between the intermediate goods. The optimality condition to the firm’s profit maximization problem then yields the demand function for intermediate inputs given by \( y_t(f) = (p_t(f)/p_t)^{-\theta} y_t \), where \( p_t = \int_0^1 y_t(f) f^{1-\theta} df \) is the price of the final good.

Following Rotemberg (1982), each firm faces a cost to adjusting its price, \( \text{adj}_t(f) \), which emphasizes the negative effect that price changes can have on customer-firm relationships. Using the functional form in Ireland (1997), \( \text{adj}_t(f) = \varphi[p_t(f)/(\pi_{t-1}(f))] - 1/2 y_t^2/p_t(t) \), the real profits of firm \( f \) are \( \delta_t(f) = (p_t(f)/p_t) y_t(f) - w_t n_t(f) - \text{adj}_t(f) \), where \( \varphi \geq 0 \) scales the size of the adjustment costs and \( \pi_t \) is the steady-state gross inflation rate. Firm \( f \) chooses its price, \( p_t(f) \), to maximize the expected discounted present value of real profits \( E_t \sum_{k=0}^{\infty} \lambda_{t,k} d_k(f) \), where \( \lambda_{t,t} \equiv 1 \), \( \lambda_{t,t+1} = \beta_{t+1} (c_{t+1}/c_t) \) is the pricing kernel between periods \( t \) and \( t+1 \) and \( \lambda_{t,k} \equiv \prod_{j=t+1}^{k} \lambda_{j-1,j} \). In a symmetric equilibrium, all firms make identical decisions and the optimality condition implies

\[
\varphi \left( \frac{\pi_t}{\pi} - 1 \right) \frac{\pi_t}{\pi} = (1 - \theta) + \theta \Psi_t + \varphi E_t \left[ \lambda_{t,t+1} (\frac{\pi_{t+1}}{\pi} - 1) \frac{\pi_{t+1} y_{t+1}}{\pi y_t} \right],
\]

(5)

where \( \Psi_t = w_t/z_t \) is the real marginal cost. In the absence of price adjustment costs (i.e., \( \varphi = 0 \), \( \Psi_t = (\theta-1)/\theta \), which is the inverse of a firm’s markup of price over marginal cost.

Each period, the central bank sets the gross nominal interest rate according to

\[
r_t = \max \{1, r^*(\pi_t/\pi^*)^{\phi_t}(y_t/y_t^*)^{\phi_y} \},
\]

(6)
where $\pi^* = \bar{\pi}$ is the inflation rate target and $\phi_{\pi}$ and $\phi_y$ are the policy responses to inflation and output. The output target is either steady-state output, $y_t^* = \bar{y}$, or potential output, $y_t^* = y_t^0 = (\chi \mu)^{-1/(1+\eta)} z_t$, which is the level of output when $\varphi = 0$. We also examine the case where $\phi_y = 0$.

The resource constraint is given by $c_t = y_t - \text{adj}_t = y_t^{adj}$, where $y_t^{adj}$ includes the value added by intermediate firms, which is their output minus quadratic price adjustment costs. A competitive equilibrium consists of sequences of quantities, $\{c_t, n_t, b_t, y_t\}_{t=0}^\infty$, prices, $\{w_t, r_t, \pi_t\}_{t=0}^\infty$, and exogenous variables, $\{\beta_t, z_t\}_{t=0}^\infty$ that satisfy the household’s and firm’s optimality conditions, (2), (3), and (5), the production function, $y_t = z_t n_t$, the monetary policy rule, (6), the stochastic processes, (1) and (4), the bond market clearing condition, $b_t = 0$, and the resource constraint.

### 2.2 Model 2: Baseline with Capital

Model 2 adds capital accumulation to Model 1. The household chooses sequences $\{c_t, i_t, n_t, b_t\}_{t=0}^\infty$ to maximize the preferences in Model 1 subject to

$$c_t + i_t + \Phi(i_t/k_{t-1})k_{t-1} + b_t = w_t n_t + r_t^k k_{t-1} + r_{t-1} b_{t-1}/\pi_t + d_t,$$

$$k_t = (1-\delta) k_{t-1} + i_t,$$

where $i_t$ is investment, $k_t$ is capital, $r_t^k$ is the real rental rate of capital, and $\Phi(\cdot)$ is a positive, increasing, and convex function that measures the cost of adjusting the capital stock. We assume $\Phi(x) = \phi(x-\delta)^2/2$, where $\phi$ controls the size of the adjustment cost. Although other papers utilize alternative specifications of capital/investment adjustment costs, we use this specification because it does not add another state variable to our model, which allows us to present the complete model solution. Optimality yields an equation for Tobin’s $q$ and a consumption Euler equation given by

$$q_t = 1 + \phi(i_t/k_{t-1} - \delta),$$

$$q_t = E_t \left[ \beta_{t+1} \frac{c_t}{c_{t+1}} \left( r_{t+1}^k - \frac{\phi}{2} \left( \frac{i_{t+1}}{k_t} - \delta \right)^2 + \phi \left( \frac{i_{t+1}}{k_t} - \delta \right) \frac{i_{t+1}}{k_t} \right) \right].$$

Intermediate firm $f \in [0, 1]$ produces a differentiated good, $y_t(f)$, according to $y_t(f) = z_t k_{t-1}^\alpha n_{t-1} (f)^{1-\alpha}$, where $k_t(f)$ and $n_t(f)$ are the levels of capital and employment used by firm $f$. Each intermediate firm then chooses its inputs to minimize operating costs, $r_t^k k_{t-1}^\alpha (f) + w_t n_t(f)$, subject to its production function, which yields a consolidated optimality condition given by

$$\alpha w_t n_t = (1-\alpha)r_t^k k_{t-1}.$$

The firm pricing equation (5) remains unchanged, except that $\Psi_t = w_t^{1-\alpha}(r_t^k)^{\alpha}/[z_t(1-\alpha)^{1-\alpha}].$

The resource constraint includes the output lost due to price and capital adjustment costs and is given by $c_t + i_t + \Phi(i_t/k_{t-1})k_{t-1} = y_t^{adj}$. A competitive equilibrium consists of sequences of quantities, $\{c_t, n_t, i_t, k_t, b_t, y_t\}_{t=0}^\infty$, prices, $\{w_t, r_t^k, r_t, \pi_t, q_t\}_{t=0}^\infty$, and exogenous variables, $\{\beta_t, z_t\}_{t=0}^\infty$ that satisfy the household’s and firm’s optimality conditions, (2), (3), (5), (9), (10), and (11), the production function, $y_t = z_t k_{t-1}^\alpha n_t^{1-\alpha}$, the monetary policy rule, (6), the stochastic processes, (1) and (4), the capital law of motion, (8), bond market clearing, $b_t = 0$, and the resource constraint.

### 3 Calibration, Solution Method, and Simulation Procedure

#### 3.1 Calibration

We calibrate the models in section 2 at a quarterly frequency using common values in the monetary policy literature. The parameters are shown in table 1. The annual real
The model is solved using the policy function iteration algorithm described in Richter et al. (2014), which is based on the theoretical work on monotone operators in Coleman (1991). This solution method discretizes the state space and uses time iteration to solve for the updated decision rules until the tolerance criterion is met. We use piecewise linear

| Frisch Elasticity of Labor Supply | $1/\eta$ | 3 | Inflation Coefficient: MP Rule | $\phi_\pi$ | 1.5 |
| Elasticity of Substitution between Goods | $\theta$ | 6 | Output Coefficient: MP Rule | $\phi_\pi$ | 0.1 |
| Rotemberg Adjustment Cost Coefficient | $\varphi$ | 59.11 | Steady-State Technology | $\bar{\pi}$ | 1 |
| Steady-State Labor | $\bar{n}$ | 0.33 | Technology Persistence | $\rho_z$ | 0.9 |
| Capital Depreciation Rate | $\delta$ | 0.025 | Technology Shock Standard Deviation | $\sigma_z$ | 0.0025 |
| Cost Share of Capital | $\alpha$ | 0.33 | Steady-State Discount Factor | $\beta$ | 0.995 |
| Capital Adjustment Cost | $\phi$ | 5.6 | Discount Factor Persistence | $\rho_{\beta}$ | 0.8 |
| Steady-State Inflation | $\pi$ | 1.006 | Discount Factor Standard Deviation | $\sigma_{\pi}$ | 0.0025 |

Table 1: Baseline calibration. A † denotes a parameter that only applies to Model 2.

The likelihood that the nominal interest rate falls to and remains at zero depends on both the parameters of the discount factor and technology processes. Richter and Throckmorton (2015) show a clear tradeoff exists between the persistence and the standard deviation of the stochastic shock processes. As the persistence of a process increases, the standard deviation of that shock must decline, otherwise our numerical algorithm will not converge to a minimum state variable (MSV) solution. The failure to converge occurs because the economy either remains at the ZLB too long when the shocks are very persistent or falls to the ZLB too frequently when the processes are highly volatile. We chose the discount factor and technology parameters so (1) They are constant across all models; (2) They generate ZLB events when simulating the model; and (3) They match the data as closely as possible. Specifically, we set the persistence of the discount factor, $\rho_{\beta}$, equal to 0.8 and the standard deviation of the shock, $\sigma_z$, equal to 0.0025. Those values follow Fernández-Villaverde et al. (2012) who assume that a discount factor shock has a half life of about 3 quarters. Steady-state technology, $\bar{n}$, is normalized to 1, the persistence of the technology shock, $\rho_z$, is 0.9, and the standard deviation of the shock, $\sigma_{\pi}$, equals 0.0025. In the data, deviations of log real GDP from trend are 1.85% per quarter and deviations of the log difference in the PCE price index are 0.29% from 1983-2007. The equivalent values in our models are smaller than is observed since additional real world shocks and sources of persistence are needed to match the data.

3.2 Solution Method The model is solved using the policy function iteration algorithm described in Richter et al. (2014), which is based on the theoretical work on monotone operators in Coleman (1991). This solution method discretizes the state space and uses time iteration to solve for the updated decision rules until the tolerance criterion is met. We use piecewise linear
interpolation to approximate future variables, since this approach more accurately captures the kink in the decision rules than continuous approximating functions, and then use Gauss-Hermite quadrature to numerically integrate. Those techniques capture the expectational effects of going to and returning to the ZLB. For a formal description of the numerical algorithm see appendix A.

Benhabib et al.’s (2001) finding that constrained New Keynesian models have two deterministic steady-state equilibria has generated considerable discussion in the literature about whether there are conditions in which a unique MSV solution exists in stochastic models with a ZLB constraint. Specifically, they find two nominal interest rate/inflation rate pairs that satisfy the steady-state equilibrium system. In one steady state, the central bank meets its positive inflation target, whereas in the other steady state the economy experiences deflation. Richter and Throckmorton (2015) show that the numerical algorithm used in our paper converges to the inflationary equilibrium as long as there is a sufficient expectation of returning to a monetary policy rule that conforms to the Taylor principle. Our algorithm, however, never converges to the deflationary equilibrium.

The intuition for how our algorithm behaves can be discerned from the simple three-equation linear New Keynesian model. We know determinacy in this model depends on whether the Taylor principle holds (i.e., the nominal interest rate moves more than one-for-one with inflation), assuming the fiscal authority ensures stable debt dynamics (i.e., passive fiscal policy). If the Taylor principle holds, our algorithm converges to the unique MSV solution that can be analytically derived. When the Taylor principle does not hold (i.e., passive monetary policy), our algorithm will not converge, even though the model has many solutions in this case. The only way our algorithm can locate these solutions is if a process for the sunspot shocks is explicitly written down.

The same rationale applies in our model with a ZLB constraint except that there are two types of sunspots. One type is analogous to the sunspots that occur when the Taylor principle does not hold. A pegged nominal interest rate is a special type of passive monetary policy, where the distribution of future shocks is truncated in a stochastic model. Thus, an occasionally binding ZLB constraint is similar to a Taylor rule that switches between an active and passive policy. As long as there is a sufficient expectation of returning to an active monetary policy, our algorithm will converge.

Davig and Leeper (2007) examine determinacy in a Fisherian economy that switches between active and passive policy. They prove that as long as one of the regimes satisfies the Taylor principle, the central bank can passively respond to inflation in the other regime and still have a determinate solution. Richter and Throckmorton (2015) show that the convergence region—the region of the parameter space where our algorithm converges to an MSV solution—is identical to the determinacy region Davig and Leeper (2007) derive. This exercise is informative because a model with an occasionally binding ZLB constraint is similar to a model with a monetary policy rule that switches between active and passive policy. Richter and Throckmorton (2015) also examine how the standard deviation of the stochastic processes affect whether the algorithm converges to the inflationary steady state in a model with a ZLB. They find that the boundary of the convergence region imposes a clear tradeoff between the expected frequency and average duration of ZLB events. Therefore, a model with a ZLB constraint produces the same intuition described in Davig and Leeper (2007). As long as the ZLB does not bind too frequently or for too long, our algorithm converges.

Wolman (2005) also uses policy function iteration to solve a New Keynesian model with a ZLB constraint. He points out that this algorithm can only locate solutions as a function of its natural state variables and is not suitable for analyzing certain types of multiplicity. He also finds that even though a deflationary steady state exists, the model may never exhibit the characteristics of that equilibrium. McCallum (2001) argues the deflationary equilibrium is not economically relevant since it is not E-stable (i.e., the economy does not converge to an equilibrium after a deviation from rational expectations beliefs). Building on that work, Christiano and Eichenbaum (2012) find evidence of multiple equilibria, including sunspots, in a nonlinear model with a ZLB constraint. They also argue that those equilibria are implausible because they are not E-stable. In our algorithm, the initial and subsequent conjectures for the decision rules deviate from the rational expectations equilibrium (REE), which is similar to learning where beliefs deviate from the REE. We also find that our algorithm only converges to the inflationary steady state.
converge to the positive inflation equilibrium. If, on the other hand, the expectation of returning to
the Taylor rule is not strong enough or the probability of returning to the ZLB is too high, then a
stable inflationary equilibrium does not exist and our algorithm will diverge. That finding does not
necessarily mean the model has no solutions. Instead, it could indicate many solutions exist, but
that finding can not be observed without specifying a process for the sunspot shocks.

The other type of sunspot shock is unique to a model with a ZLB constraint. The existence
of both an inflationary and a deflationary steady state means the economy could fluctuate between
them. Therefore, we could add a Markov chain to our existing models that governs switches
between the two steady states as in Aruoba and Schorfheide (2013). Appendix B shows the con-
vergence properties of our algorithm with a series of numerical exercises. We first replicate the
multiple deterministic equilibria result in Benhabib et al. (2001) and then study three versions of
Model 1: a perfect foresight version, a version with a stochastic discount factor process, and a ver-
sion with a 2-state Markov chain governing switches between the two deterministic steady states.
In each case, the algorithm converges to a solution around the inflationary steady state.

Uncertainty continues to exist about whether these sunspot shocks affect an economy with a
ZLB constraint. Economists, for example, want to understand if the sunspot shocks are observed
in the data or even reflect dynamics that are economically feasible. Although addressing those
questions is important for future research, our analysis, like most macroeconomic research on the
ZLB constraint, is concentrated on examining solutions around the inflationary steady state.

3.3 Simulation Procedure  We simulate the models using draws from the distributions for
the discount factor and technology shocks. Figure 2 plots the distributions of the state variables
and the nominal interest rate in a 500,000 quarter simulation of Model 1. The vertical axes show
the frequency of each realization as a percent of the simulation length. Variables on the horizontal
axes are shown as percent deviations from steady state, except the nominal interest rate which
is a net percentage. The dashed lines represent the bounds of the state space, which are chosen
to minimize extrapolation of the decision rules in the simulation. The solid lines denote the
theoretical unconditional distributions scaled for comparison with the simulated distributions.

Figure 2a shows the unconditional distributions of technology, the discount factor, and the
nominal interest rate. The state space for technology lies within ±2.5% of its steady state, which
is normalized to unity. The state space for the discount factor lies between ±1.9% of its steady
state, which equals 0.995. Across these states, the quarterly net nominal interest rate is distributed
over a range of 0% to 3.6%, with a large mass (5% of quarters) between 0 and 20 basis points. As
we demonstrate below, model dynamics are very different when the policy rate lies in this interval.

Figure 2b shows the distribution of the discount factor and technology conditional on the ZLB
binding. A high discount factor is the primary source of ZLB events, as indicated by the difference
between its distribution conditional on the ZLB (bars) and its theoretical unconditional distribution
(solid line). The conditional distribution for the discount factor is centered around 1% above steady
state. A higher discount factor means households are more willing to postpone their consumption.

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8We fix the bounds of the state space prior to solving the model. For the exogenous state variables, we know
how wide to set the grids to guarantee minimal extrapolation when simulating the model. For capital, which is an
endogenous state variable, we first solve the model and then check that the bounds on capital are wide enough to
eliminate extrapolation. We resolve the model with wider grids until there is no extrapolation in our simulation.

9In all of our results, a hat denotes percent deviation from the deterministic steady state (i.e., for some generic
variable \( x \) in levels, \( \hat{x}_t = 100(x_t - \bar{x})/\bar{x} \) and a tilde denotes a net rate (i.e., for some gross rate \( x \), \( \tilde{x}_t = 100(x_t - 1) \)).
Lower consumption pushes down inflation, which in turn causes the nominal interest rate to fall. If households are patient enough, then the nominal interest rate hits its ZLB. The nominal interest rate can also fall to zero when technology is sufficiently far above its steady state because higher supply leads to lower prices. Our result is consistent with the finding in Fernández-Villaverde et al. (2012) that high levels of technology are associated with a low nominal interest rate.

The Fed’s policy rate has been at its effective ZLB since December 2008. Gust et al. (2013) show that most financial market participants expected the federal funds rate to remain below 25 basis points for only a few quarters. For example, the median forecast in the first quarter of 2009 was below 25 basis points only until the third quarter of that year and gradually increased to 2.5% in 2012. The Survey of Professional Forecasters (SPF) conducted by the Philadelphia Fed asks its participants to forecast the 1-year T-Bill rate up to four quarters in the future. The median (solid line) and 16/84 percentiles (dashed lines) of the individual forecasts in the first quarter of 2009 are shown in figure 3. The median forecast predicted the T-Bill rate would exceed 50 basis points while the 84th percentile predicted it would hit 1% within 1 year. Those forecasts indicate that people expected the ZLB to bind for just a few quarters even though the recession was quite severe.

Our model is calibrated to the average time the ZLB is expected to hold and not to the duration of the current ZLB episode in the U.S. With that being said, it is possible for longer ZLB events

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Figure 2: Model $1 (y_t^* = \bar{y})$ distributions as a percentage of a 500,000 quarter simulation. Each variable is in percent deviations from its steady-state value. The dashed lines are the bounds of the state space. The solid lines are the theoretical unconditional distributions of the state variables scaled for comparison with the conditional distributions.
to occur in our framework. Figure 4, for example, shows the distribution of the length of each ZLB event as a percentage of the total number of those events in a 500,000 quarter simulation of Model 1 \( y^*_t = \bar{y} \). The vertical dashed line indicates the average ZLB duration is 1.87 quarters. The longest ZLB event is 19 quarters, which is about the length of the current ZLB episode. ZLB events with a duration of 1, 2, and 3 quarters account for 58.4%, 21.2%, and 9.5%, respectively, of all ZLB events in the simulation. Therefore, our calibration of the stochastic processes produces a distribution of ZLB event durations that is similar to household expectations at the onset of the Great Recession. The calibration for Model 2 also yields a similar distribution of ZLB events.

4 MODEL 1: STATES OF THE ECONOMY, ECONOMIC DYNAMICS, AND THE ZLB

The New Keynesian model without capital, outlined in section 2.1, contains two state variables, the discount factor and technology. This section presents the complete solution to Model 1, key cross sections of that solution, impulse responses to technology shocks, and simulation statistics. We compare these results across alternative monetary policy rules. Each variable is shown in percent deviations from its steady state, except inflation and the interest rates, which are net percentages.

Figure 5 shows three-dimensional contour plots of the net nominal interest rate and adjusted output over the entire state space. These plots provide a complete picture of the model solution for both variables when the central bank targets steady-state output \( y^*_t = \bar{y} \). The shaded areas represent the states of the economy where the net nominal interest rate, \( \tilde{r} \), equals zero. Those areas reveal the nominal interest rate only hits the ZLB when either technology or the discount factor are unusually high. When the central bank targets steady-state output, a higher level of technology lowers inflation and the real interest rate when the ZLB does not bind. When the ZLB binds, higher technology continues to push down inflation, which forces up the real interest rate and causes demand to fall. Looking at the highest discount factor in figure 5, output exhibits the same unconventional response, even when technology is at or below its steady state. In fact, many studies assume an elevated discount factor is the cause of the current ZLB event in the U.S.

The contours in figure 5 are useful because they provide the solution for every combination of
Figure 5: Model 1 \( \left( y^*_{t} = \bar{y} \right) \) decision rules as a function of the technology \( \left( \hat{z}_{-1} \right) \) and the discount factor \( \left( \hat{\beta}_{-1} \right) \) states. Each variable is in percent deviations from its deterministic steady state, except the nominal interest rate, which is a net percentage. The shaded region indicates where the ZLB binds.

The two shocks, but they can be difficult to read. Therefore, we focus on specific cross sections of the state space. The solid line in Figure 5 shows the cross section where the technology state is held constant at its steady state \( \left( \hat{z}_{-1} = 0 \right) \). Two-dimensional representations of that cross section are shown in Figure 6. The shaded region highlights where the ZLB binds, which begins when the discount factor is 0.9% above its steady state. A high discount factor indicates that households have a strong desire to save. Elevated savings depresses demand, which reduces output, inflation, and the nominal interest rate. At the ZLB, any further reduction in expected inflation is offset by an equal increase in the real interest rate. That higher real interest rate raises the cost of current consumption which further lowers demand in discount factor states where the ZLB binds.

The dashed line in Figure 5 shows the cross section where the discount factor is held constant at 0.9% above its steady-state value \( \left( \hat{\beta}_{-1} = 0.9 \right) \), which is the minimum value where the ZLB binds when technology is at its steady state. Figure 7 shows a two-dimensional representation of that cross section and the same cross sections for different values of \( \phi_y \). The darkest shaded region indicates where the ZLB binds when \( \hat{z}_{-1} = 0 \) and \( \phi_y = 0.1 \). Smaller values of \( \phi_y \) cause the ZLB to first bind in slightly lower technology states, as the lighter shaded regions show. The unconventional response of the economy to a positive technology shock is smaller as the value of \( \phi_y \) declines. With \( \phi_y = 0.05 \left( \phi_y = 0 \right) \), the response of output is positive in technology states up to 0.67% (1.3%) above its steady state. Furthermore, in high technology states where the economy does contract, output and inflation are more stable with a lower \( \phi_y \). For example, when \( \phi_y = 0 \), output never falls below its initial ZLB level \( \left( \hat{y}^{adj} = -1.34\% \right) \), except in the highest technology states. In contrast, when \( \phi_y = 0.1 \), output falls from -1.18% to -3.34% when technology increases from \( \hat{z}_{-1} = 0 \) to \( \hat{z}_{-1} = 2.5 \). It is clear from those results that a shorter expected duration at the ZLB can reverse the unconventional dynamics, since the expected duration of the ZLB increases in higher technology states. That finding is consistent with the conclusions of Braun and Körber (2011). It is also apparent that the monetary policy rule plays an important role in the dynamics at the ZLB.
since the slopes of the decision rules differ greatly across the alternative values of $\phi_y$.

To better understand our results, we begin by examining the region of the state space where the ZLB does not bind. In low technology states, workers are less productive and firms’ per unit marginal cost of production is higher. Firms respond by raising prices and reducing their demand for labor. With less output available for consumption, the household wants to work more to moderate the decline in consumption. The higher labor supply dominates the drop in labor demand, so the equilibrium level of labor is higher and the real wage rate is lower. The household also believes technology will slowly return to its steady state and as a result, expects its future consumption to increase. Higher expected future consumption is reflected in an elevated real interest rate. A larger value of $\phi_y$ in technology states where the ZLB does not bind keeps output, labor, and the real wage rate closer to their steady states, but that additional stability comes at the expense of more inflation and a higher nominal interest rate. The real interest rate in that case is mostly unaffected.

The last area to consider are the technology states where the ZLB binds. In those states, higher technology continues to lower per unit production costs and firms react by lowering their prices. The additional decline in expected inflation when the nominal interest rate equals zero raises the real interest rate. The household reduces its consumption and increases its labor supply to capitalize on the higher returns which results in the paradox of thrift. Aggregate demand falls because everyone wants to save more at the higher real interest rate, but that is not possible in equilibrium. Thus, the lower demand reduces output until actual and desired savings are equal. Firms respond to the decrease in demand by further lowering prices and cutting labor demand. The drop in labor demand dominates the increase in labor supply, so that both total hours and the real wage decline. This is an example of the paradox of toil [Eggertsson (2010)]. At the ZLB, everyone wants to work
more, but the higher real interest rate lowers demand, which causes firms to reduce employment.\footnote{The standard deviation of the stochastic processes affects the expected frequency and average duration of the ZLB. Appendix C shows that the qualitative effects of a larger $\phi_y$ in Figure 7 are similar when $\rho_\beta = 0.75$.}

With a smaller response to the deviations from steady-state output, inflation is more stable in all technology states. Thus, the real interest rate rises less at the ZLB, which helps maintain household demand in high technology states. Higher labor demand raises equilibrium hours, which mitigates the decline in the real wage. In short, a tension exists at the ZLB between the supply-side effects of technology and the demand-side effects of the real interest rate. If the central bank responds less aggressively to the deviations from steady-state output when the ZLB does not bind, then the

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Figure 7: Model 1 (\(y_t^* = \bar{y}\)) decision rules as a function of the technology state (\(\hat{z}_{-1}\)). The discount factor state (\(\hat{\beta}_{-1}\)) is fixed at the minimum value that causes the ZLB to bind when \(\hat{z}_{-1} = 0\) and \(\phi_y = 0.10\). Each variable is in percent deviations from its deterministic steady state, except inflation and the interest rates, which are net percentages. The shaded region indicates where the ZLB binds for a given \(\phi_y\) value.
demand-side effects at the ZLB are weaker and both real and nominal variables are less volatile.

We also examine the effects of technology shocks by computing generalized impulse response functions (GIRFs) of a policy shock. GIRFs provide a clear quantitative comparison between economic dynamics at and away from the ZLB. They are based on an average of model simulations where the realization of shocks is consistent with the household’s expectations over time. Figure 8 plots the generalized impulse responses to a 1% positive technology shock when the central bank targets steady-state output under two sets of initial conditions: (1) a non-ZLB case (solid line), where the discount factor remains at its steady state so that the nominal interest rate is above its ZLB; and (2) a ZLB case (dashed line), where the discount factor is set to its mean value over a 500,000 quarter simulation under the condition that the ZLB binds and technology is at its steady state. To compute the GIRFs, we calculate a baseline path as the mean of 10,000 simulations of the model conditional on only the initial state vector. We then calculate a second mean from another set of 10,000 simulations, but in this case the shock in the first quarter is replaced with a 1 standard deviation positive technology shock. We compute the percentage change (or difference for the interest rates and inflation) between the two means. Those values are shown on the vertical axis.

The impulse responses in the non-ZLB case are standard and follow the intuition from the decision rules. On average, a 1% positive technology shock increases adjusted output, lowers firms’ per unit marginal cost of production, and causes inflation and the nominal interest rate to fall. According to the Taylor rule, the nominal interest rate falls more than the inflation rate, so the real interest rate declines, which increases consumption. The positive technology shock also raises productivity, which decreases the equilibrium level of labor and increases the real wage rate.

In the ZLB case, a 1% positive technology shock initially increases adjusted output by only 0.05% on average. That sluggish response occurs because the ZLB binds in 87% of the simulations after the positive technology shock, which means the nominal interest rate cannot fall by as much as it does in the non-ZLB case. The positive technology shock also lowers per unit production costs which helps to push down prices. With prices falling and the nominal interest rate stuck at zero, an increase in technology sharply raises the real interest rate. That spike then limits the increase in output and causes labor to fall further than in the non-ZLB case. Our results in figure 7, however, indicate that the responses of output and the real wage depend critically on the value of $\phi_y$. If $\phi_y = 0$, a positive technology shock increases adjusted output more on impact than when $\phi_y > 0$ because the absence of a policy response to output limits the upward pressure on the real interest rate at the ZLB. From period 2 onward, adjusted output increases as the economy exits the ZLB due to the mean reversion in both technology and the discount factor. The nominal interest rate rises far enough above zero by period 8 that the ZLB case effectively mirrors the non-ZLB case. In both cases, technology returns to its steady state about 20 quarters after the initial shock.

Figure 9 plots the same cross section of the state space that is shown by the dashed line in figure 5 across three monetary policy rules: (1) The central bank does not respond to output ($\phi_y = 0$, solid line); (2) The central bank targets steady-state output ($y^*_t = \bar{y}, \phi_y = 0.1$, dashed line); and (3) The central bank targets potential output ($y^*_t = y^*_t, \phi_y = 0.1$, circle markers). The shaded region indicates where the ZLB binds, but the level of technology where that occurs depends on the policy rule. When $\phi_y = 0 (y^*_t = y^*_t)$, the ZLB first binds when technology is 0.26% (0.08%) below its steady state. The most noteworthy difference among the policy rules is that higher technology states at the ZLB generate further increases in output and the real wage rate when the central bank

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11 The general procedure for computing GIRFs is outlined in Koop et al. (1996). See appendix D for details.
Figure 8: Model 1 ($y^*_t = \bar{y}$) GIRFs to a 1% positive technology shock. The steady-state case (solid line) is initialized at the model’s steady state. The ZLB case (dashed line) is initialized at the average state vector conditional on the ZLB binding in a 500,000 quarter simulation.

Potential output rises and falls with technology while steady-state output remains unchanged. When technology is below (above) its steady state, potential output is lower (higher) than steady-state output. A positive (negative) technology shock generates the largest positive (negative) output gap when the central bank targets steady-state output. That response raises the volatility of inflation so inflation is less stable with a steady-state output target than a potential output target.

When technology is above steady state, it lowers inflation which causes the real interest rate to rise at the ZLB. That higher real rate encourages households to reduce demand and save more. Lower demand dampens the upward pressure on output from the decline in production costs. Which effect dominates depends on whether the real interest rate rises enough to offset the positive effects of higher technology. Given that the real interest rate is inversely related to the expected inflation rate at the ZLB, the real interest rate rises less when the central bank targets potential output since inflation is more stable. Therefore, output is higher and the economy will exit the
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Figure 9: Model 1 decision rules as a function of the technology state ($\hat{z}_{-1}$). The discount factor is fixed at the minimum value that causes the ZLB to bind when $\hat{z}_{-1} = 0$ and $\phi_y = 0$. On the solid line, the central bank does not respond to output ($\phi_y = 0$). On the dashed line, it responds to deviations from steady-state output ($\hat{y}_t^* = \bar{y}$) and on the line with circle markers it responds to deviations from potential output ($\hat{y}_t^* = y_n$). Each variable is in percent deviations from its deterministic steady state, except inflation and the interest rates, which are net percentages. The shaded region indicates where the ZLB binds for a given specification of monetary policy.

ZLB quicker when the central bank targets potential output instead of steady-state output.

Next, we examine how the output target affects the likelihood of hitting the ZLB and the volatility of output and inflation using 500,000 quarter simulations. Table 2a shows the effect of reducing the weight on output ($\phi_y$) while holding the weight on inflation at $\phi_\pi = 1.5$. We begin with the value in Taylor (1993), $\phi_y = 0.125$, and reduce it by increments of 0.025. With a steady-state output target ($\hat{y}_t^* = \bar{y}$), the ZLB binds in 2.73% of the simulated quarters and has an average duration of 1.90 quarters when $\phi_y = 0.125$. These values monotonically decrease with $\phi_y$ and equal 2.33% and 1.81 quarters when $\phi_y = 0$. Decreasing the weight on the steady-state output target raises the
volatility of output but has non-monotonic effects on the volatility of inflation. Technology shocks generate negative co-movement between inflation and output, whereas discount factor shocks produce positive co-movement. Therefore, an increase in φ_y induces competing effects on inflation volatility. Discount factor shocks dominate at low values of φ_y (≤ 0.10) so inflation variability declines. When φ_y ≥ 0.125, technology shocks dominate which raises the volatility of inflation.\(^{12}\)

The results are reversed when the central bank targets potential output (\(y_t^n = \bar{y}_t^n\)). Placing more weight on the deviations from potential output reduces the likelihood of hitting the ZLB and the volatility of both output and inflation. Those results are consistent with the findings in the optimal policy literature. Adam and Billi (2006) setup a New Keynesian model where the central bank minimizes the expected value of a loss function that depends on inflation and the potential output gap. They show it is optimal for the central bank to aggressively reduce its policy rate after an adverse shock when it faces a ZLB constraint because it reduces visits to the ZLB.

<table>
<thead>
<tr>
<th>φ_y</th>
<th>ZLB Binds</th>
<th>Avg. ZLB Quarters</th>
<th>Std. Dev. (% of mean)</th>
<th>ZLB Binds</th>
<th>Avg. ZLB Quarters</th>
<th>Std. Dev. (% of mean)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>% of quarters</td>
<td>Output Inflation</td>
<td>% of quarters</td>
<td>Output Inflation</td>
<td>% of quarters</td>
<td>Output Inflation</td>
</tr>
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<td>1.90</td>
<td>0.6501</td>
<td>0.3326</td>
<td>1.56</td>
<td>1.72</td>
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<td>0.6704</td>
<td>0.3308</td>
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<td>0.3311</td>
<td>1.80</td>
<td>1.75</td>
</tr>
<tr>
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<td>0.7167</td>
<td>0.3335</td>
<td>1.95</td>
<td>1.77</td>
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<td>0.7719</td>
<td>0.3447</td>
<td>2.33</td>
<td>1.81</td>
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(a) Volatility implications of alternative weights on the output gap (\(φ_y\)). The weight on inflation is \(φ_π = 1.5\).

<table>
<thead>
<tr>
<th>φ_π</th>
<th>ZLB Binds</th>
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<th>Std. Dev. (% of mean)</th>
<th>ZLB Binds</th>
<th>Avg. ZLB Quarters</th>
<th>Std. Dev. (% of mean)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>% of quarters</td>
<td>Output Inflation</td>
<td>% of quarters</td>
<td>Output Inflation</td>
<td>% of quarters</td>
<td>Output Inflation</td>
</tr>
<tr>
<td>1.500</td>
<td>2.73</td>
<td>1.90</td>
<td>0.6501</td>
<td>0.3326</td>
<td>1.56</td>
<td>1.72</td>
</tr>
<tr>
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<td>1.72</td>
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<td>0.2582</td>
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</table>

(b) Volatility implications of alternative weights on the inflation gap (\(φ_π\)). The weight on the output gap is \(φ_y = 0.125\).

Table 2: Model 1: No capital, technology and discount factor shocks.

Table 2b reports the results when we fix \(φ_y = 0.125\) and change the response to the deviations of inflation from its target (\(φ_π\)). With \(y_t^n = \bar{y}_t^n\), the probability of hitting the ZLB falls from 2.73% of the simulated quarters in the baseline case to 0.43% when \(φ_π = 3\). In addition, the standard deviations of output and inflation fall as \(φ_π\) increases. A higher \(φ_π\) reduces the volatility of output and inflation and decreases the likelihood of hitting the ZLB when \(y_t^n = \bar{y}_t^n\). In the baseline case, \(12\)In figure 7, the shaded region shrinks as φ_y declines, which suggests that the ZLB is less likely to bind, contrary to table 2a. That perceived contradiction occurs because as φ_y decreases the ZLB region twists, meaning in some states of the economy the ZLB binds when it did not with a higher φ_y, whereas in others it no longer binds. States of the economy where the ZLB binds are less likely to occur when φ_y is low, since they correspond, on average, to more extreme realizations of the shocks. Thus, the percentage of quarters in which the ZLB binds increases with φ_y.\(^{12}\)
the ZLB binds in 1.56% of the quarters and 0.38% when $\phi_n = 3$. Overall, table 2 demonstrates that the central bank’s response to both output and inflation as well as the specification of the output gap affects the simulation properties of the model—both qualitatively and quantitatively.

5 Model 2: States of the Economy and the ZLB

This section shows how our findings change when capital is incorporated into a New Keynesian model. In Model 1, the household can only smooth consumption by varying its labor supply. The presence of capital in Model 2 gives the household another margin to smooth consumption. The addition of another state variable, however, complicates the presentation of the complete solution to the model. We initially fix technology at its steady state, so the complete solution can be presented with contour plots. That allows us to focus on the dynamics created by the discount factor process, which is commonly used to generate ZLB events in the literature. Thus, this model initially contains two state variables—the discount factor and the endogenous capital stock. We then reintroduce the technology process to compare the dynamics between Models 1 and 2 in response to a technology shock. Also in this section, we focus on the dynamics when the central bank targets steady-state output since we believe it better reflects the behavior of actual monetary policy.\(^{13}\)

Figure 10 shows the three-dimensional contour plots of the nominal interest rate, output, consumption, and investment over the entire state space. Presenting a complete picture of the solution is informative in models with an endogenous state variable like capital since it shows the interaction between the two states. The curvature of the ZLB (shaded) region is due to the quadratic capital adjustment costs. When capital is at its steady state ($\hat{k}_{-1} = 0$), the ZLB binds when $\hat{\beta}_{-1} = 1.22$. As capital rises, the nominal interest rate initially hits zero at lower values of the discount factor. In general, the qualitative behavior of consumption, inflation, and the nominal interest rate are similar to the model without capital. The household’s ability to invest in capital, however, causes consumption to be less volatile and generates stronger expectational effects of the ZLB.

We focus our analysis on two cross sections of the contour map in figure 10. The endogeneity of capital makes selecting particular cross sections in Model 2 more difficult than in Model 1. In Model 1, the discount factor and technology states are independent; therefore, any one realization of the discount factor is just as likely regardless of the technology state. In Model 2, the capital and discount factor states are not independent, so the level of capital is likely below (above) its steady-state value when the discount factor is also below (above) its steady-state value.

Figure 11 shows two cross sections from the contour map in figure 10. The solid line is the cross section where the capital is fixed at its steady state ($\hat{k}_{-1} = 0$). The dashed line represents the cross section where capital increases with the discount factor along the diagonal of the state space ($\hat{k}_{-1} = \hat{k}_{\text{diag}}$). The darker (entire) shaded region indicates the area of the state space where the ZLB binds in the steady-state (diagonal) cross section. We begin by examining the behavior of the economy when the ZLB does not bind. Regardless of the capital state, a higher discount factor makes the household more patient, which increases their desire to invest in capital and to postpone consumption. The higher discount factor also encourages the household to supply more labor. The additional investment raises the marginal product of labor, which causes firms to raise their output and labor demand. The increase in output leads to lower inflation. In equilibrium, labor increases

\(^{13}\)Potential output is defined as the level of output under flexible prices. In Model 1, potential output is exogenous and we can solve for it analytically, but in Model 2 it has no closed-form solution. It is impossible to numerically solve a flexible price model with a ZLB because sticky prices are necessary for an equilibrium to exist in our model.
and the real wage rate falls as the discount factor increases. Finally, a higher capital stock pushes down its marginal product which lowers the real rental rate in elevated discount factor states.

In the diagonal cross section where capital increases with the discount factor ($\hat{k}_{-1} = \hat{k}_{\text{diag}}$), the marginal product of capital falls as the discount factor rises which leads to a more rapid decline in the real rental rate. From the household’s perspective, a lower rental rate makes investment less attractive as a consumption smoothing channel. The household responds by moderating both their decline in consumption and their increase in labor supply compared to the case in which capital is fixed at its steady state ($\hat{k}_{-1} = 0$). Those more modest responses in the real rental rate, investment, consumption, and labor are illustrated by their flatter decision rules when the ZLB does not bind.

In the diagonal (steady-state) cross section, the ZLB binds when the discount factor is more than 0.8% (1.2%) above its steady state. The qualitative properties of the decision rules when the
Figure 11: Model 2 ($y^*_{it} = \bar{y}$) decision rules as a function of the discount factor state ($\hat{\beta}_{-1}$). The solid line is the cross section of the state space where the capital state is fixed at its steady-state value ($\hat{k}_{-1} = 0$), and the dashed line is the diagonal cross section where the capital state changes with the discount factor state ($\hat{k}_{-1} = \hat{k}_{diag}$). Each variable is in percent deviations from its deterministic steady state, except inflation and the nominal interest rate, which are net percentages. The dark (entire) shaded region indicates where the ZLB binds when $\hat{k}_{-1} = 0$ ($\hat{k}_{-1} = \hat{k}_{diag}$).
ZLB binds are similar across both cross sections. The mechanism that distorts the economy in Model 2 is similar to Model 1. As the discount factor rises, both inflation and the real interest rate continue to fall. When the nominal rate hits zero, the real interest rate rises as inflation continues to fall. That higher real rate further encourages households to postpone consumption and motivates them to supply more labor. Firms respond to the lower demand by further reducing their prices and sharply cutting their labor demand. That decline in labor demand dominates the increase in labor supply so that both the equilibrium level of labor and the real wage rate fall. Lower consumption then pushes down output, which causes the household to reduce investment even more in order to further smooth its consumption. Thus, the paradoxes of toil and thrift both occur—despite the household wanting to work more to smooth consumption and save more to benefit from higher real interest rates, both hours and investment fall. Those findings demonstrate that our model with capital produces the same unconventional dynamics as the model without capital.

The Importance of Nonlinearities

We apply our policy function iteration algorithm to log-linearized versions of Model 1 and Model 2, where the only nonlinearity is the ZLB constraint. This solution method is similar to the procedure employed in Nakov (2008), where linear splines are used to approximate the kink in the decision rules. We then compare the resulting linear decision rules to their nonlinear counterparts to demonstrate the importance of using the fully nonlinear model. The benefit of solving the nonlinear and linear models in this manner is that differences in the decision rules are entirely due to whether or not the model is linearized.

Figure 12 compares cross sections of the linear and nonlinear decision rules for output in Model 1 when the central bank targets steady-state output. The left (right) panel shows the decision rule as a function of the technology state (discount factor state). The linear decision rules are a fairly accurate approximation of the nonlinear decision rules for both the technology and the discount factor shocks so long as the ZLB does not bind. The linear and nonlinear decision rules for output then diverge as the economy moves deeper into the ZLB region for both shocks. That separation is initiated by the inability of the central bank to compensate for growing price adjustment costs, which are different due to the linearization of the quadratic price adjustment cost function. Furthermore, the location of the ZLB kink is nearly identical for both the linear and nonlinear decision rules. Those results, in contrast to Fernández-Villaverde et al. (2012), indicate that the linear model provides a fairly good approximation of the nonlinear model without capital in most states.

Figure 13 compares the linear and nonlinear decision rules in Model 2 as a function of the discount factor state. The ZLB first binds in the linear (nonlinear) model when the discount factor is 1.4% (0.8%) percent above its steady state. That difference has two important implications. One, simulations of the linear model indicate that the economy hits the ZLB far less frequently than in the nonlinear model. In our baseline calibration, the ZLB never binds in the linear model, while it binds in 1.15% of the time in the nonlinear model. That result is one reason why ZLB studies that linearize the model must specify much larger shocks to generate ZLB events. Two, the decision rules differ between the linear and nonlinear model when the ZLB does not bind, because the expectational effects of visiting the ZLB are weaker in the linear model. Thus, the linear model cannot accurately quantify the effects of discount factor shocks even when the ZLB does not bind. The reason the linear and nonlinear models generate such different results in the model with capital is because Model 2 permits asset substitution in contrast to Model 1. As the discount factor rises,

14 The rental rate of capital falls at the ZLB, but the household expects that the future rental rate will increase since they believe the discount factor will return to its steady state. That result is consistent with a rising real interest rate.
Figure 12: Model 1 \((y^*_t = \tilde{y})\) decision rules as a function of the technology state (left panel) and the discount factor state (right panel). The solid line (dashed line) corresponds to the decision rules based on the constrained nonlinear (linear) model. Each variable is in percent deviations from its deterministic steady state. The dark (entire) shaded region indicates where the ZLB binds in the fully nonlinear (linear) model.

Figure 13: Model 2 \((y^*_t = \tilde{y})\) decision rules as a function of the discount factor state \((\hat{\beta}_{-1})\). The capital state changes with the discount factor state \((\hat{k}_{-1} = \hat{k}_{diag})\). The solid line (dashed line) corresponds to the decision rules based on the constrained nonlinear (linear) model. Each variable is in percent deviations from its deterministic steady state, except inflation and the interest rates, which are net percentage. The dark (entire) shaded region indicates where the ZLB binds in the fully nonlinear (linear) model.
households increase investment, which reduces the marginal product of capital and the real rental rate. Forward looking households expect a much lower rental rate at the ZLB, which reduces real and nominal interest rates before the ZLB is hit. Since the consequences of the ZLB are more severe in the nonlinear model, the expectational effects of the ZLB are also stronger.

6 Model 1 and Model 2 Comparisons

This section shows capital qualitatively and quantitatively affects dynamics at the ZLB. We compare GIRFs in our models with and without capital because cross sections of the decision rules require assumptions about how the capital state in Model 2 co-moves with the exogenous state variables. To conduct such an experiment, technology is stochastic with the same parameter values in both models. We also assume that the central bank targets steady-state output \( (y^*_t = \bar{y}) \) and set \( \phi_y = 0.025 \) in both models. A small weight on \( \phi_y \) is necessary to make a direct comparison because our numerical algorithm does not converge for higher values of \( \phi_y \) in Model 2.

![Figure 14](image)

Figure 14: Comparison of the GIRFs to a 1% positive technology shock in Model 1 (solid line) and in Model 2 (dashed line). The initial state vector is equal to the average state vector in a 500,000 quarter simulation conditional on the ZLB binding. The central bank targets steady-state output \( (y^*_t = \bar{y}) \).

Figure 14 plots the responses to a 1% positive technology shock in Model 1 (solid line) and Model 2 (dashed line). The discount factor is initially set to its mean value that causes the ZLB to bind in a 500,000 quarter simulation of the model where technology shocks are set to zero (Model 1: \( \hat{\beta}_{-1} = 1 \) and Model 2: \( \hat{\beta}_{-1} = 1.4 \)). The unconventional dynamics are not present in Model 1 when \( \phi_y \) is small because the positive supply-side effects of higher technology dominate the negative demand-side effects of a higher real interest rate. As \( \phi_y \) increases, the adverse demand-side effects overcome the beneficial supply-side effects so that output and labor hours both decline in response to a positive technology shock at the ZLB. Output, in contrast, declines on impact in Model 2 even when \( \phi_y \) is small. The responses of the real interest rate, inflation, and labor are
qualitatively the same in both models but quantitatively larger in Model 2. From period 2 onward, technology and the discount factor mean revert and the responses from both models converge.

Next, we show the impact of capital on the volatility of output and inflation for a range of values for \( \phi_y \). Model 1 and Model 2 are simulated for 500,000 quarters under the assumption that the central bank targets steady-state output. We fix technology at its steady state to examine a broader range of values for \( \phi_y \).

Table 3 shows the effect of reducing the weight on the output gap (\( \phi_y \)) while holding the weight on inflation (\( \phi_\pi \)) at 1.5. The value of \( \phi_y \) is initially set slightly below the original Taylor (1993) specification, \( \phi_y = 0.1 \), and is then reduced by increments of 0.025.

<table>
<thead>
<tr>
<th>( \phi_y )</th>
<th>Model 1</th>
<th></th>
<th>Model 2</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>% of quarters</td>
<td>ZLB Binds</td>
<td>Avg. ZLB</td>
<td>Std. Dev. (% of mean)</td>
<td>ZLB Binds</td>
</tr>
<tr>
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<td>1.63</td>
<td>0.4972</td>
<td>0.2769</td>
</tr>
<tr>
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<td>1.29</td>
<td>1.64</td>
<td>0.5168</td>
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<tr>
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<td>1.65</td>
<td>0.5382</td>
<td>0.2997</td>
</tr>
<tr>
<td>0.025</td>
<td>1.51</td>
<td>1.66</td>
<td>0.5615</td>
<td>0.3126</td>
</tr>
<tr>
<td>0.000</td>
<td>1.64</td>
<td>1.68</td>
<td>0.5870</td>
<td>0.3268</td>
</tr>
</tbody>
</table>

Table 3: Volatility implications of alternative weights on the output gap. Comparison between Model 1 and Model 2. The only stochastic component in both models is discount factor shocks. \( \phi_\pi = 1.50 \), \( \rho_\beta = 0.80 \), and \( \sigma_\beta = 0.0025 \).

Model 2 generates two qualitatively different results from Model 1. One, both output and inflation volatility decline as \( \phi_y \) increases in Model 1, whereas a tradeoff exists between lower output volatility and higher inflation volatility in Model 2. Since capital introduces another channel through which households smooth consumption, an increase in the discount factor raises investment and reduces demand for consumption goods. The net effect on output is positive so that discount factor shocks cause inflation and output to move in opposite directions unlike Model 1. Thus, there is a tradeoff between inflation and output stability in Model 2 that is similar to the tradeoff the central bank faces in response to technology shocks in Model 1. Two, the likelihood of the ZLB increases with higher values of \( \phi_y \) in Model 2, whereas it decreases in Model 1.\(^{15}\) That difference is a consequence of more inflation volatility with higher values of \( \phi_y \) in Model 2.

Capital is essential to explain business cycles, but it is largely absent from work on the ZLB. Including capital in the model changes the qualitative effects of demand shocks and alters the effects of a monetary policy rule that emphasizes output stability. Our results in this section highlight the importance of including capital when analyzing the impact of monetary policy at the ZLB.

7 Conclusion

This paper examines monetary policy when the nominal interest rate is constrained by the ZLB using models with and without capital. We use these models to analyze why technology shocks at the ZLB may have unconventional effects, what factors influence the likelihood of hitting the ZLB, and the implications of alternative monetary policy rules. Four main findings emerge: (1) A positive technology shock can generate lower consumption, labor, and output when the ZLB binds

\(^{15}\)Table 3 appears to contradict our Model 1 findings when the central bank targets steady-state output (see table 2a, col. 2 and 3). Since technology is fixed, steady-state and potential output are equal. Thus, the qualitative result is consistent with the findings from Model 1 when the central bank targets potential output (see table 2a, col. 6 and 7).
and the central bank targets steady-state output. Those unconventional dynamics usually disappear when the central bank targets potential output; (2) When the central bank targets steady-state output in the model with capital, a positive technology shock at the ZLB generates more contractionary dynamics than in the model without capital; (3) The presence of capital changes the qualitative effects of demand shocks and alters the impact of a monetary policy rule that emphasizes output stability; and (4) The constrained linear model provides a good approximation of the nonlinear model without capital, but differences exist between the solutions in the model with capital.

In spite of the amount of research on the ZLB, many important questions remain. For example, do the medium- to long-run benefits of returning to normal policy outweigh the short-run costs of a higher policy rate? What are the benefits of forward guidance and quantitative easing in a dynamic model that accounts for expectational effects? Lastly, how does the impact of discount factor and technology shocks change as a result of those policies? This paper provides a careful discussion of the dynamics in the types of models that researchers may use to answer these questions.

REFERENCES


——— (2010): “On the Size of the Fiscal Multiplier When the Nominal Interest Rate is Zero,” Mimeo, University of Tokyo.


A Numerical Algorithm

A formal description of the numerical algorithm begins by writing the model compactly as

$$E[f(v_{t+1}, w_{t+1}, v_t, w_t)|\Omega_t] = 0,$$

where $f$ is vector-valued function that contains the equilibrium system, $v$ is a vector of exogenous variables, $w$ is a vector of endogenous variables, and $\Omega_t = \{M, P, z_t\}$ is the household’s information set in period $t$, which contains the structural model, $M$, its parameters, $P$, and the state vector, $z$. In Model 1, $v = z = (\beta, z)$ and $w = (c, \pi, y, n, w, r)$. In Model 2 with both the technology and discount factor processes, $v = (\beta, z)$, $z = (k, \beta, z)$, and $w = (c, \pi, y, n, w, r, k, i, r^k, q)$.

The algorithm approximates the vector of decision rules, $\Phi$, as a function of the state vector, $z$. We iterate on $\Phi = (c, \pi)$ for Model 1 and $\Phi = (n, \pi, i)$ for Model 2 so that we can easily solve for future variables that enter the household’s expectations using $f$. Each continuous state variable in $z$ is discretized into $N^d$ points, where $d \in \{1, \ldots, D\}$ and $D$ is the dimension of the state space. The discretized state space is represented by a set of unique $D$-dimensional coordinates (nodes). We set the bounds of the exogenous state variables to encompass 99.999% of the probability mass of the distribution. For capital, which is an endogenous state variable, we first solve the model and then check that the bounds on capital are wide enough to eliminate extrapolation. We resolve the model with wider grids until there is no extrapolation in our simulation. We specify 101 grid points for each state variable and use 31 Gauss-Hermite weights for each shock. Those techniques minimize extrapolation and ensure that the location of the kink in the decision rules is accurate.

The following outline summarizes the policy function algorithm we employ for our models. Let $i \in \{0, \ldots, I\}$ index the iterations of the algorithm and $n \in \{1, \ldots, \prod_{d=1}^D N^d\}$ index the nodes.
1. Obtain initial conjectures for the decision rules on each node from the log-linear model without a ZLB constraint. The initial conjectures are \( \hat{c}_0 \) and \( \hat{n}_0 \) for Model 1 and \( \hat{n}_0, \hat{\pi}_0 \), and \( \hat{i}_0 \) for Model 2. We use Sims (2002) gensys.m program to obtain these conjectures.

2. For \( i \in \{1, \ldots, I\} \), implement the following steps:

   (a) On each node, solve for \( \{y_t, r_t\} \) given \( \hat{c}_{i-1}(z^n_t) \) and \( \hat{\pi}_{i-1}(z^n_t) \) in Model 1 and given \( \hat{n}_{i-1}(z^n_t), \hat{\pi}_{i-1}(z^n_t), \) and \( \hat{i}_{i-1}(z^n_t) \) in Model 2 with the ZLB constraint imposed.

   (b) Linearly interpolate \( (c_{t+1}, \pi_{t+1}) \) in Model 1 and \( (n_{t+1}, \pi_{t+1}, i_{t+1}) \) in Model 2 given \( \{\varepsilon^m_{t+1}\}_{m=1}^M \). Each of the \( M \) values \( \varepsilon^m_{t+1} \) are Gauss-Hermite quadrature nodes. We use Gauss-Hermite quadrature to numerically integrate, since it is accurate for normally distributed shocks. We use piecewise linear interpolation to approximate future variables, since this approach more accurately captures the kink in the decision rules than continuous approximating functions such as cubic splines or Chebyshev polynomials.\(^{16}\)

   (c) We use the nonlinear solver, csolve.m, to minimize the Euler equation errors. On each node, numerically integrate to approximate the expectation operators,

   \[
   \mathbb{E} \left[ f(x^m_{t+1}, x^n_t) | \Omega_t \right] \approx \frac{1}{\pi} \sum_{m=1}^{M} f(x^m_{t+1}, x^n_t) \phi(\varepsilon^m_{t+1}),
   \]

   where \( x \equiv (v, w) \) and \( \phi \) are the respective Gauss-Hermite weights. The superscripts on \( x \) indicate which realizations of the state variables are used to compute expectations. The nonlinear solver searches for \( \hat{c}_i(z^n_t) \) and \( \hat{\pi}_i(z^n_t) \) in Model 1 and \( \hat{n}_i(z^n_t), \hat{\pi}_i(z^n_t), \) and \( \hat{i}_i(z^n_t) \) in Model 2 so that the Euler equation errors are less than \( 10^{-4} \) on each node.

3. Define \( \text{maxdist}_i \equiv \max\{|\hat{c}_i - \hat{c}_{i-1}|, |\hat{\pi}_i - \hat{\pi}_{i-1}|\} \) in Model 1 and \( \text{maxdist}_j \equiv \max\{|\hat{n}_i - \hat{n}_{i-1}|, |\hat{\pi}_i - \hat{\pi}_{i-1}|, |\hat{i}_i - \hat{i}_{i-1}|\} \) in Model 2. Repeat step 2 until one of the following occurs:

   - If for all \( n \), \( \text{maxdist}_i < 10^{-13} \) for 10 consecutive iterations, then the algorithm converged to a MSV solution. In Model 1, since the state is composed of only exogenous variables, the solution is bounded so long as the decisions rules are positive and finite. In Model 2, simulations of the model must not be explosive.
   - Otherwise, we say the algorithm is non-convergent for one of the following reasons:
     - \( i = I = 500,000 \) (Algorithm times out)
     - For all \( n \) and any \( i \), \( \hat{\pi}_i < 0.5 \), or for any \( n, \hat{c}_i < 0 \) in Model 1 or \( \hat{n}_i < 0 \) in Model 2 (Approximating functions drift)
     - Define \( \text{dir}_i = \text{maxdist}_i - \text{maxdist}_{i-1} \). For all \( n \), \( \text{dir}_i \geq 0 \) and \( \text{dir}_i - \text{dir}_{i-1} \geq 0 \) for 50 consecutive iterations (Algorithm diverges).

   The same criteria is used to generate the results in Richter and Throckmorton (2015).

\(^{16}\)Aruoba and Schorfheide (2013) use a linear combination of two Chebyshev polynomials—one that captures the dynamics when the ZLB binds and one that captures the dynamics when the Taylor principle holds. This approach is more accurate than using one Chebyshev polynomial, but there is no guarantee that it will accurately locate the kink. Moreover, Chebyshev polynomials can lead to large approximation errors due to extrapolation. With linear interpolation, a dense state space will lead to more predictable extrapolation and more accurately locate the kink. See Richter et al. (2014) for a comparison of these two solution methods in a New Keynesian model with a ZLB constraint.
B CONVERGENCE PATHS AND STEADY-STATE EQUILIBRIA

Figure 15 illustrates that the model has two steady states, which is consistent with Benhabib et al. (2001). The left panel highlights the intersections of the consumption Euler equation and the monetary policy rule in steady state (circles). Those two steady states are

\[ \bar{r} = \frac{\bar{\pi}}{\bar{\beta}}, \]  

(Consumption Euler Equation)

\[ \bar{r} = \max \{1, r^*(\bar{\pi}/\pi^*)^{\phi_{\pi}} \}, \]  

(Interest Rate Rule)

which results in two steady-state inflation rates:

\[ \bar{\pi} = \begin{cases} 
\pi^* & \text{when } \bar{r} = r^* \\
\frac{\bar{\pi}}{\bar{\beta}} & \text{when } \bar{r} = 1 
\end{cases}. \]

When combined with the first-order condition for labor and the resource constraint, the firm pricing equation yields the steady-state value of consumption as a function of the steady-state inflation rate:

\[ \hat{c} = \left( \frac{1}{\theta \chi} \left( (1 - \beta) \varphi \left( \frac{\bar{\pi}}{\pi^*} - 1 \right) \frac{\bar{\pi}}{\pi^*} - (1 - \theta) \right) \left( 1 - \frac{\varphi}{2} \left( \frac{\bar{\pi}}{\pi^*} - 1 \right)^2 \right)^{\eta} \right)^{1/(1 + \eta)}. \] (12)

Since the model contains two steady-state inflation rates, consumption also has two steady-state values, which are shown in the right panel of figure 15. In this section, the inflation rate is shown as a net percentage and consumption is in percent deviations from its positive inflation steady state.

![Figure 15](image-url)  

Figure 15: Model 1 steady states (circles). The left panel shows the consumption Euler equation (black line) and the interest rate rule (dashed line) in steady state. The right panel shows the firm pricing equation in steady state.

Figure 16 shows the convergence paths of consumption and inflation when our algorithm is initialized at different points (diamonds and crosses). The convergence paths correspond to the values of consumption and inflation in each iteration of our numerical algorithm. The left panel shows the model’s two steady states (circles). Inflation is positive in one steady state and negative in the other steady state. The right panel shows our model may converge to the deflationary steady state via a saddle path that runs from the northwest and southeast as shown in Benhabib et al.
Figure 16: Convergence paths to the steady-state equilibria (circles) in the deterministic version of Model 1. A diamond denotes an initial conjecture that converges to the positive inflation steady state, and a cross denotes an initial conjecture that asymptotically converges to a corner solution where there is no consumption.

That saddle path is not explicitly shown in the right panel because the algorithm only converges to the deflationary steady state if the initial conjecture is exactly equal to one of the values on either side of the stable manifold. Obtaining that precise convergence path is not possible without the analytical equation for the saddle path because any numerical algorithm is based on an approximation of the true solution. Thus, our algorithm only converges to the deflationary steady state if the distance between the initial conjecture and the deflationary steady state is less than the tolerance criterion, $10^{-10}$.

The unstable manifolds point away from the deflationary steady state toward the southwest. Our algorithm converges to the inflationary steady state so long as the initial conjectures for inflation and consumption are to the northeast (diamonds) of the stable manifolds of the deflationary steady state. Initial conjectures in the southwest (crosses) yield paths that are unstable because they asymptotically approach a corner solution where consumption is equal to 0.

Figure 17 shows the convergence paths for consumption and inflation when the discount factor state, $\beta_{-1} \neq \bar{\beta}$. The right (left) panel displays the convergence paths for $\beta_{-1} = 0.9975$ ($\beta_{-1} = 0.9925$), which is above (below) $\bar{\beta} = 0.995$. In both cases, the inflationary steady state is stable, but there is no evidence the algorithm converges to the deflationary steady state as in figure 16. Furthermore, our findings do not indicate that there is a saddle path to the deflationary steady state.

Figure 18 compares the convergence paths of consumption and inflation when $\beta_{-1} = \bar{\beta}$ for the deterministic (left panel) and stochastic (right panel) versions of Model 1. In the stochastic model, the household forms expectations over future realizations of $\beta$. The paths in the stochastic model differ from the deterministic model in two important ways. One, fewer initial conjectures converge to the inflationary steady state in the stochastic model. That decline occurs because expectations are formed over future values of $\beta$ that are in the region where the paths of consumption and inflation diverge from the inflationary steady state. In other words, additional instability is created.

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$^{17}$Christiano and Eichenbaum (2012) describe the deflationary steady state as a pencil standing on its tip. If the agent’s belief is incorrect by $10^{-9}$, then the equilibrium falls apart (i.e., the pencil falls over). Initial conjectures for consumption and inflation that slightly deviate from the deflationary steady state mean the algorithm either converges to the inflationary steady state or diverges away from the deflationary steady state.
by forming expectations over values of $\beta$ that put the algorithm on an unstable path. Two, the deflationary steady state is no longer present in the stochastic model. If the stochastic model is initialized at values of consumption and inflation on either stable manifold in the deterministic model, then many realizations of $\beta$ result in paths that diverge from the inflationary steady state.

To further understand these convergence paths, we also examine a Markov-switching specification of the stochastic model [Eggertsson and Woodford (2003)]. This exercise demonstrates that expectational effects in the stochastic model destabilize the deflationary steady state that is present in the deterministic model. In this model, the equilibrium at time $t$ is determined by a 2-state Markov chain with transition matrix $Pr\{s_t = j|s_{t-1} = i\} = p_{ij}, i, j \in \{1, 2\}$. The two equilibria

---

**Figure 17:** Convergence paths to the steady-state equilibria (circles) in the deterministic version of Model 1. A diamond denotes an initial conjecture that converges to the positive inflation steady state, and a cross denotes an initial conjecture that asymptotically converges to a corner solution where there is no consumption.

**Figure 18:** Convergence paths to steady states (circles) for the perfect foresight and stochastic models when $\beta_{-1} = \bar{\beta}$. A diamond denotes an initial conjecture that converges to the positive inflation steady state, and a cross denotes an initial conjecture that asymptotically converges to a corner solution where there is no consumption.
are the inflationary and deflationary steady states of the deterministic model shown in figure 15:

\[
(\bar{\pi}, \bar{r}) = \begin{cases} 
(\pi^*, r^*) & \text{for } s_t = 1 \\
(\bar{\beta}, 1) & \text{for } s_t = 2
\end{cases}
\]

Figure 19 shows the paths of consumption and inflation beginning from their initial conjectures, \((\hat{c}_0, \hat{\pi}_0)\), for each state, \(s\). In both panels, the inflationary steady state is perfectly absorbing, \(p_{11} = 1\), which is often assumed in the ZLB literature [e.g., Braun et al. (2013); Christiano et al. (2011); Eggertsson and Woodford (2003)]. The left panel shows that when the initial conjecture for \(s = 2\) is equal to the deflationary steady state and its state is perfectly absorbing (i.e., \(p_{22} = 1\)) the algorithm converges to the deflationary steady state. The right panel, however, shows that if \(p_{22} = 0.99\) (i.e., there is a small probability of leaving the deflationary steady state), then only the inflationary steady state remains. Conditional on starting in state 2, even the smallest possibility of returning to the inflationary steady state makes the deflationary steady state unstable. The deflationary equilibrium satisfies the steady-state system of equations of the deterministic model, but it no longer satisfies the stochastic and dynamic system of equations due to expectational effects.

![Figure 19: Convergence paths in a version of Model 1 that switches between the steady states. A diamond denotes an initial conjecture that converges to a steady state. \(s = 1\) (\(s = 2\)) is the inflationary (deflationary) steady state.](image)

C Sensitivity Analysis

The parameters of the stochastic processes impact where the ZLB binds in the state space, the slope of the decision rules, and model dynamics. As an example, figure 20 compares the Model 1 decision rules with \(\rho_\beta = 0.8\) and \(\rho_\beta = 0.75\) when technology is fixed at its steady-state. A more persistent discount factor process makes the ZLB bind in lower discount factor states. When \(\rho_\beta = 0.8\), the nominal interest rate is stuck at its ZLB whenever \(\hat{\beta}_{-1} > 0.9\), whereas when \(\rho_\beta = 0.75\), the ZLB does not bind unless \(\hat{\beta}_{-1} > 1.3\). This result means households will expect that the nominal interest rate will hit its ZLB less frequently and in situations when it does hit its ZLB, they will expect it to bind for fewer quarters. When \(\rho_\beta = 0.75\), the average ZLB event is only 1.3 quarters,
Figure 20: Model 1 ($y^* = \bar{y}$) decision rules as a function of the discount factor state ($\hat{\beta}_t$) when the persistence is high (solid line) and low (dashed line). The technology state is fixed at its steady-state value ($\hat{z}_t = 0$). Each variable is in percent deviations from its deterministic steady state, except inflation and the interest rates, which are net percentages. The light (dark) shaded region indicates where the ZLB binds when $\rho_\beta = 0.8$ ($\rho_\beta = 0.75$).

Figure 21: Model 1 ($y^* = \bar{y}$) decision rules as a function of the technology state ($\hat{z}_t$). The discount factor persistence is $\rho_\beta = 0.75$. The discount factor state ($\hat{\beta}_t$) is fixed at the minimum value that causes the ZLB to bind when $\hat{z}_t = 0$ and $\phi_y = 0$. Each variable is in percent deviations from its deterministic steady state, except inflation and the interest rates, which are net percentages. The shaded region indicates where the ZLB binds for a given $\phi_y$ value.
compared to 1.9 quarters when $\rho_\beta = 0.8$. In states of the economy where the ZLB does not bind, the dynamics are virtually identical to the dynamics that occur when $\rho_\beta = 0.8$. At the ZLB, the economy is far more sensitive to shocks that affect demand because the central bank can no longer blunt the effects of those shocks by lowering the nominal interest rate. Thus, the slopes of the decision rules for inflation and output are steeper and the real interest rate rises sharply at the ZLB.

Figure 21 reproduces figure 7 when $\rho_\beta = 0.75$. These results are informative because they show that increasing the weight on the steady-state output target does not affect our qualitative results. The decline in output at the ZLB for a given value of $\phi_\pi$ is less severe because the household expects that the nominal interest rate will rise in the near future and that the central bank will be able to stabilize output and inflation. A higher value of $\phi_y$, however, reduces the positive effect of technology shocks on output and in unusually high technology states, these shocks can reduce output. Those findings are consistent with our results in section 4. If we increase the value of $\rho_\beta$, the unconventional effects of positive technology shocks at the ZLB are even more pronounced.

D  GENERALIZED IMPULSE RESPONSE FUNCTIONS

The GIRFs are based on an average of 10,000 simulations of the model. The advantage of GIRFs is that the realizations of the shocks are consistent with the household’s expectation that the stochastic processes will mean revert when the model is initialized away from its stochastic steady state. The procedure for calculating GIRFs is described in Koop et al. (1996). We apply the following steps:

1. Find the state vector at which to initialize each case:
   a. **Non-ZLB Case**: Simulate the model without shocks until it converges to its stochastic steady state, $z_{0+}$.
   b. **ZLB Case**: Simulate the model for 500,000 quarters using random draws of discount factor shocks. The initial state vector is the average state vector conditional on the ZLB binding, $z_{0lb}$. The average discount factor when the ZLB binds is $1\%$ above steady state.

2. Draw random shocks, $\{\varepsilon_{x,t}, \varepsilon_{\beta,t}\}_{t=0}^{N}$, from their independent normal distributions. Simulate each case for $R$ different draws of the sequence of shocks beginning at the alternative initial state vectors, $z_{0+}^s$ and $z_{0+}^b$. This process generates $R$ equilibrium paths for each case, $\{x_i^j(z_{0+}^s)\}_{t=0}^{N}$ and $\{x_i^j(z_{0+}^b)\}_{t=0}^{N}$, where $j \in \{1, 2, \ldots, R\}$. We set $N = 20$ and $R = 10,000$.

3. Using the same $R$ draws of shocks from step 2, replace the technology shock in period 1 with a 1% shock (i.e., set $\varepsilon_{z,1} = 0.01$ for all $j \in \{1, 2, \ldots, R\}$). Then re-simulate the model to obtain $R$ equilibrium paths for each case, $\{x_i^j(z_{0+}^s, \varepsilon_{z,1})\}_{t=0}^{N}$ and $\{x_i^j(z_{0+}^b, \varepsilon_{z,1})\}_{t=0}^{N}$.

4. Average across the $R$ simulations from step 2 and step 3 to obtain average paths given by
   $$\bar{x}_t(z_{0+}^s) = \frac{1}{R} \sum_{j=1}^{R} x_i^j(z_{0+}^s), \quad \bar{x}_t(z_{0+}^s, \varepsilon_{z,1}) = \frac{1}{R} \sum_{j=1}^{R} x_i^j(z_{0+}^s, \varepsilon_{z,1}),$$
   $$\bar{x}_t(z_{0+}^b) = \frac{1}{R} \sum_{j=1}^{R} x_i^j(z_{0+}^b), \quad \bar{x}_t(z_{0+}^b, \varepsilon_{z,1}) = \frac{1}{R} \sum_{j=1}^{R} x_i^j(z_{0+}^b, \varepsilon_{z,1}).$$

5. The difference between the two average paths for each case is a GIRF. In our figures, a variable in either case with a hat is calculated as $100(\bar{x}_t(z_{0}, \varepsilon_{z,1})/\bar{x}_t(z_{0}) - 1)$ and with a tilde is calculated as $100(\tilde{x}_t(z_{0}, \varepsilon_{z,1}) - \tilde{x}_t(z_{0}))$, where $s \in \{ss, zlb\}$. 35