

What Helps Forecast U.S. Inflation?—Mind the Gap!*

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Abstract

The Phillips curve, which posits a relationship between inflation and domestic economic activity, introduces a crucial trade-off between real and nominal objectives for the central bank. Atkeson and Ohanian (2001), among others, present evidence that forecasts of U.S. inflation from Phillips curve-based models tend to underperform relative to naïve forecasts. We propose that globalization can be an important factor in explaining the poor performance of forecasts under a closed-economy Phillips curve. To illustrate that, we empirically evaluate the performance of open-economy Phillips curve-based forecasts constructed with global variables, such as G7 credit growth, G7 money supply growth, terms of trade, and the real effective exchange rate. These global variables perform significantly better than domestic variables, and serve as proxies for poorly-measured indicators of global slack. Moreover, we show that forecasts using simulated data from a workhorse open-economy New Keynesian model support our empirical findings on the open-economy Phillips curve and also suggest that better monetary policy and aspects of the Great Moderation have improved the forecast accuracy of open-economy models.

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1 Introduction

Forecasting inflation—accurately and reliably—plays a critical role for policy-making and for the decisions of the private sector in making long-term nominal commitments. In macroeconomic analysis and inflation forecasting, the traditional Phillips curve has been a widely used model that broadly captures the empirical relationship between inflation and unemployment rate, capacity utilization or output gap.

As documented by Atkeson and Ohanian (2001), the Phillips curve has flattened since 1984. Their finding was that the Phillips curve-based models did not yield more accurate forecasts than the naïve, 4 quarter random walk benchmark. Stock and Watson (2007) emphasized the role of lower volatility in inflation in the U.S. and in the world during this period. The risk of naïve forecasts, computed as the mean squared forecast error (MSFE), declined as well. Forecasts under a Phillips curve specification have become, in turn, less accurate by the MSFE metric. A survey by Stock and Watson (2008) suggests recent forecasts based on univariate specifications, including the Phillips curve, only occasionally performed well.¹ The literature starting with Atkeson and Ohanian (2001) has set a key threshold in forecasting inflation using Phillips curves in reduced form. In particular, it has become important to understand how Phillips curves can yield accurate forecasts that outperform naïve specifications.

A prominent explanation to the break in the Phillips curve that is suggested in the literature is globalization—the integration of global markets in goods, labor, and capital. The recent literature postulates the ‘global slack hypothesis’, i.e. the hypothesis that foreign slack, as well as domestic slack, drives domestic inflation in the short-run, as a way to reconcile the Phillips curve relationship with the evidence in an increasingly more integrated world. Hence, the open-economy Phillips curve that ties inflation to global measures of economic activity has become a focus of investigation.

The nexus between globalization, inflation, and monetary policy has been partly discussed in academic and policy circles. Bernanke (2007), González-Páramo (2008), Mishkin (2007), Papademos (2007), Rogoff (2006), and Weber (2007) argued that globalization might have altered the inflation process, and moreover, monetary policy decisions might be affected by global economic conditions particularly in the short and medium runs.² Related theoretical and empirical literature provides inconclusive evidence on the global slack hypothesis. For instance, Binyamini and Razin (2007), and Martínez-García and Wynne (2010) make theoretical explanations and Borio and Filardo (2007) provides empirical evidence for the global slack hypothesis. On the other hand, Ball (2006), Ihrig et al. (2007), Pain et al. (2006) find empirical results that do not support the global slack hypothesis. Milani (2010) and Milani (2012), among others, argue that the foreign economic activity has a role on domestic supply and demand, but its effect on domestic inflation is negligible, finding weak evidence for the global slack hypothesis.

In this paper, we evaluate the open-economy Phillips curves both theoretically and empirically. We suggest that in a workhorse open-economy New Keynesian (NK) model, an open-economy Phillips curve should be a relevant specification for forecasting inflation, but why it may not be a successful specification in practice is largely an issue that can be explained by imperfectly-measured output gap variables used in

¹Poor performance in inflation forecasts is not limited to the statistical literature. Edge and Gürkaynak (2010), for example, reported an analogous result on the performance of inflation forecasts based on a medium-scale Dynamic Stochastic General Equilibrium (DSGE) framework during this period.

²It is important to note that while globalization might have challenged central banks with more complicated macroeconomic issues compared to the case of a closed economy, its impact on inflation should nevertheless be viewed as a complementary channel to monetary policy, rather than an alternative. See also Fisher (2005), Fisher (2006), and Woodford (2010) for further discussion on globalization and monetary policy.

these forecasts. Hence, even when the theoretical validity of an open-economy Phillips curve is assured, teasing out empirical support from the data and forecasting inflation under the open-economy framework is a challenging task. It is, in general, difficult to find sufficiently long, reliable and robust time series of global slack—global output gap or capacity utilization. Quality and data availability concerns are discussed in the current paper as well as in previous studies. Therefore, it becomes particularly useful to find accurate proxies for global slack—and we show that global money growth, credit growth, terms of trade and REER can be used as proxies for global slack.

Undoubtedly, it is important to understand the theoretical basis for why a given variable would be useful for forecasting. In this regard, we first try to explain how money growth might help forecast inflation. It might be appealing to relate the performance of money growth to the quantity theory of money, but would the quantity theory of money really help us understand money growth as a useful variable to forecast inflation? Woodford (2008) argues that even strong empirical evidence in the long run relation between money growth and inflation does not necessarily imply that money growth will be useful for forecasting inflation. The main reason, as he explains, is that cointegration of money growth with the inflation rate would imply that in order to forecast the average inflation rate in the long run, it would be sufficient to know what the average growth rate of money will be in the long run; and therefore no other variables would be necessary to forecast the average inflation rate over the same horizon. However, one does not know the long run average money growth rate, as it is an endogenous variable with respect to the central bank's policy. If it were an exogenous variable, then it could make sense to detect long run trends from the moving averages of recent observations. Woodford (2008) also presents an example with a simple NK model where, in theory, money growth will not outperform slack even in the long run.³

Even though the quantity theory of money, and the empirical evidence for it, may not imply that money has actual value in forecasting inflation, we believe that the empirical success of money measures is quite obvious in our findings and should not be ignored. Moreover, the primary mechanism underlying the high performance of money growth observed in our study, as well as in D'Agostino and Surico (2009), may be related to a different channel than the one posited by the conventional approach to the quantity theory of money. We consider the following alternative argument in the NK tradition. In the context of the NK model, money moves as a result of changes in endogenous variables that are themselves related to the output gap. As Woodford (2008) also argues, money demand is a residual in this model, however, it is one that we can measure more easily and that could still provide a signal for the output gap fluctuations that we can use to forecast future inflation. For instance, if we can say that in the context of this model the money growth in equilibrium is proportional to output gap, then money growth is a signal for a measure of slack that we actually do not observe properly (i.e. the output gap) and as such can still be exploited to forecast inflation. If money is tied to slack in equilibrium, then it might be used as a variable instead of slack which is easily defined in theory but cannot be measured perfectly in practice.

In a recent work, Martínez-García and Wynne (2010) suggested a similar role for terms of trade. They showed that under a variant of the open-economy NK framework and in a standard representation of the open-economy Phillips curve, global slack can be replaced with domestic slack and the terms of trade gap. Hence, the information content of terms of trade can be exploited to forecast domestic inflation without

³It might be arguable whether money growth will not outperform slack, but the statement is valid as far as a standard New Keynesian model is considered. Woodford (2008) also suggests that in theory, the usefulness of money growth depends on the framework one considers.

using the global slack, which is difficult to measure in practice. In principle, the particular open-economy NK model we study implies that the real exchange rate is proportional to the terms of trade, so the forecasting performance of both variables should be the same. In more general models, the proportionality breaks down, but presumably we would still have a significant correlation between terms of trade and the REER that would explain if one variable is helpful to forecast inflation, the other one may be helpful, too. The similarity in the performances of terms of trade and REER confirms that both measures can help forecast inflation.⁴

The credit growth measures that we evaluate have not been tested in the literature so far. Moreover, it is hard to understand, in theory, why it may work as a good forecasting variable, as standard models do not incorporate credit. Stock and Watson (1999b) evaluated the performances of some credit measures, which are either subcomponents of the monetary aggregates (like the monetary base or reserves) or related to commercial and industrial loans. The latter group includes the truly credit-related variables in their analysis, but they are quite different than our measures of credit. Stock and Watson (1999b)'s focus on credit to firms is indeed sensible since credit to firms should be related with the tone of economic activity and the costs that the firms face. However, they do not find evidence that such a channel works very well to help forecast inflation. We look at credit to the overall private sector, including households, and our findings show that it matters for forecasting inflation. The intuition behind this would simply be that what really matters is the money that reaches the hands of consumers and allows them to finance purchases of goods. Putting more money in the hands of consumers chasing the same goods will surely put upward pressure on prices. We believe this explains why our results are so strong with credit to the private sector. However, a more detailed analysis still needs to be done and we leave this as an open question for future research.

Foreign factors for U.S. inflation can be seen in both the important role that terms of trade or REER, as well as global money or credit growth, can play in predicting inflation. Both types of variables can be very useful as proxies for the imperfectly measured global slack, at least within the context of the open-economy NK model. Even if some of these variables have been used in testing their predictive performances for U.S. inflation, the interpretation that we provide is novel. Our ultimate argument is that these measures can be helpful to forecast inflation because they could be proxies for the unobserved global slack. Our theoretical and empirical evidence suggests that global slack, in fact, has been an important factor driving the dynamics of inflation in the U.S., which the literature based on closed-economy Phillips curves has not recognized yet.

The proportionality of the global slack and macroeconomic measures discussed in the open-economy NK framework might strengthen or weaken depending on changes in the model's parameters and therefore a realistic calibration of the parameters is needed to understand to see whether these theoretical linkages are quantitatively important. In particular, our strategy is to use a model that can capture the effects of two other complementary hypotheses in addition to globalization—good luck and improved monetary policy—that are commonly discussed in the literature as plausible explanations for the observed strengths and weaknesses in the forecasting performances. To this end, we simulate data based on the model and use the data to conduct forecasts similar to those in the empirical section. We estimate mean square forecast errors

⁴Stock and Watson (1999b) and Stock and Watson (2008) evaluate the predictive ability of (nominal) U.S. trade-weighted effective exchange rate as well as a set of foreign exchange rates (Stock and Watson (1999b)). They find that these variables do not improve upon the autoregressive process of inflation or the Phillips curve-based forecasts, unlike what our findings suggest.

(MSFEs) for many plausible parameter values that capture changes in trade openness, volatility in productivity or monetary policy shocks (which we call ‘good luck’) and effectiveness of monetary policy reflected in Taylor rule parameters. For most of these patterns of forecast accuracy, we find that greater openness has improved the performance of global macroeconomic variables, while increased anti-inflationary monetary policy and the decline in the volatility of productivity shocks seem to have improved the performance of both domestic and global measures.

In regards to our experiments, Inoue and Rossi (2011) come closest to ours in that it takes account of the parameter instability. They have a richer large-scale model and a recursive procedure that tests for parameter instability of individual parameters allowing for other parameters to be unstable at the same time. They argue that the literature, in general, follows a one-at-a-time analysis as a result of which they are more likely to conclude that the only plausible channel might be good luck because all other alternative explanations tend to cancel out among themselves. Our strategy builds on this idea and, therefore, explores the impact of instability in multiple parameters at a time. We estimate the key parameters independently but not through the full model and we find accordingly that some of these structural parameters indeed matter more than previously acknowledged.

Our empirical investigation of open-economy Phillips curve forecasts starts with testing whether a set of global slack measures have predictive power for U.S. inflation. These measures are constructed by mostly theoretically-consistent output gap or capacity utilization series of the U.S. and several different groups of countries combined.

Our first finding is that one should really ‘mind the gap’. Perhaps in agreement with the existing empirical literature, these global slack variables yield mixed results in predicting different inflation measures. In fact, we confirm a result that is also found in the literature that studies the closed-economy Phillips curve-based forecasts. In theory, there should be no other current period variables that help us to forecast future inflation that can outperform the forecasts attained with the current period output gap (and lagged inflation).⁵ In practice, the difficulty of estimating the potential output and therefore deriving the output gap, makes slack measures less reliable. Therefore, it becomes key to consider other variables that can proxy for it in forecasting inflation and are more accurately measured in the data.

We then move on to an extensive evaluation of the predictive performances of a set of macroeconomic variables that can substitute global slack in forecasting inflation, as we show in our theoretical analysis. In particular, we test domestic and global money supply growth,⁶ domestic and global credit growth, as well as variables tied to the open-economy Phillips curve such as terms of trade, real effective exchange rates (REER) and domestic and global output gap. The striking finding is that all of these variables help improve upon the naïve forecasts obtained with the specification of a simple autoregressive process for inflation. We find particularly strong results for the late 1980s and onwards, the period of break in the Phillips curve pointed out by Atkeson and Ohanian (2001). The results are robust to various forecast horizons, inflation measures and estimation samples.

Our benchmark estimation and forecast periods are 1980:Q1-1991:Q4 and 1992:Q1-2011:Q4, respectively. The goal of these forecasts is to consider a wide range of relatively short series of slack⁷ and other macro-

⁵See Woodford (2008), and particularly an example on p. 1583 which shows that in a standard New Keynesian model this is the case. The paper, among others, also points out to the problem of overestimating the potential output, as a pitfall in Phillips curve-based policy making.

⁶See, for example, D’Agostino and Surico (2009) on the forecasting value of money growth.

⁷Our domestic slack measures include the CBO, FRBD, OECD, IMF U.S. slack measures as well as HP-filtered U.S. real GDP. Our

economic variables,⁸ and compare their performances. We then look into some selected variables where we go back as far as 1947:Q1 and perform rolling window forecasts, to the extent that data series are available. Specifically, we test the predictive performances of: a domestic slack series (CBO U.S. slack), two global slack series (OECD Total slack and FRBD G7 slack), a measure of domestic liquidity growth (U.S. M2 growth) and global liquidity growth (G7 average of monetary aggregates), a measure of domestic credit growth (U.S. non-financial lending from all sectors to the private sector), a measure of global credit growth (G7 average), and two measures of terms of trade (terms of trade and terms of trade ex. oil, filtered), and U.S. trade-weighted real effective exchange rate (REER, filtered).

We conduct pseudo out-of-sample forecasts for eight measures of U.S. inflation at horizons varying between 1-quarter to 12-quarters ahead. In particular, we use CPI, core CPI (all items ex. food and energy), PCE deflator, core PCE (all items ex. food and energy), GDP deflator, PPI, Sticky Prices (Sticky CPI), and Sticky Prices ex. shelter (Sticky core CPI).⁹

Our metric for forecast accuracy is the MSFE of a reduced form of a new open-economy Phillips curve with distributed lags of inflation and any of the measures of slack (or proxies for it) that we investigate, relative to the MSFE of the ‘restricted’ forecast derived from a univariate, autoregressive process of inflation, hence we consider nested models. We compute bootstrap standard errors for the MSFEs following Clark and McCracken (2006). We report the following stylized facts:

- Forecasts with the domestic slack measure perform significantly better than the simple AR process of inflation until late 1980s and particularly at short horizons. The global slack measures only outperform the simple AR process significantly in the late 1980s and at short horizons.
- Forecasts using terms of trade measures perform significantly better relative to the naïve forecasts, during the late 1980s. However, we observe switches in the performances of the two measures; in general, terms of trade ex. oil performs well whenever terms of trade does not and vice versa. Due to limited data availability, we cannot observe the performances of terms of trade ex. oil during the 1960s, but this is a period where terms of trade significantly outperforms the naïve forecasts followed by a break during late 1970s and early 1980s.
- Forecasts with REER exhibit similar patterns with terms of trade: based on the relatively short time series we observe that they perform well after the mid-1980s.
- Forecasts with money exhibit a highly significant and accurate pattern and, in general, global money growth outperforms its domestic counterpart. With an exception in PPI inflation and for recent years in the GDP deflator, the results are robust to all inflation measures and especially over medium-to-long horizons.
- Both domestic and global measures of credit growth are very successful variables in forecasting U.S. inflation in the late 1980s. Global credit growth exhibits an even more accurate path in forecasting inflation than the U.S. credit growth.

global slack measures are FRBD G7, FRBD G39, OECD G7, OECD Total and IMF Advanced. More details on the data can be found in Appendix E.

⁸In particular, we use measures of domestic and global money growth, terms of trade, and REER.

⁹We use these two sticky price measures only in the benchmark forecasts.

These extensive forecasting exercises reveal that, especially for the period starting after the break in the Phillips curve, several measures of economic activity help improve upon forecasts with the autoregressive process of inflation. These measures also outperform the closed economy Phillips curve-based forecasts. While the global counterparts of the money, credit and slack variables seem to be more useful for forecasting than the domestic measures, it is also interesting that other variables regarding the global economic activity, such as terms of trade and REER yield highly accurate forecasts compared to the Phillips curve-based forecasts.¹⁰

2 Insights from Theory

In this section, we briefly describe the building blocks of the workhorse open-economy NK model in Martínez-García and Wynne (2010) which is a variant of the model of Clarida et al. (2002). The model maintains the assumption of a zero-inflation steady state, but is augmented to incorporate a time-varying inflation target for the central bank in the short-run. Since the setup of the model we use is otherwise extensively discussed in Martínez-García and Wynne (2010) and Appendix B, here we shall put the emphasis on the key equations of its log-linearized representation and their economic interpretation.

At the core of this model we encounter an explicit open-economy version of the Phillips curve relating domestic inflation and global slack together. First, we show the key theoretical relationships relevant for forecasting inflation implied by the model. We illustrate how the closed-economy Phillips curve is not the appropriate modelling specification for forecasting domestic inflation whenever the economy is integrated with the rest of the world through trade. Second, based on simulated forecasts from the model, we investigate the role of a number of competing hypothesis—good luck, changes in the conduct of monetary policy and greater openness to trade—to explain a potential decline in the domestic inflation forecasting performance of traditional Phillips-curve-based forecasting models that rely solely on domestic slack.

2.1 The Workhorse Open-Economy New Keynesian Model

There are two countries, *Home* and *Foreign*, of equal size. The core structure of the model consists of three log-linearized equations for each country and two fundamental exogenous shocks (productivity shocks and monetary shocks). That system of equations fully characterizes the dynamics of aggregate output, inflation, and the short-term nominal interest rate in both the *Home* and *Foreign* countries.

All other endogenous variables which describe the aggregate behavior of the economy in each country can be expressed as linear functions of the two fundamental shocks, aggregate output, inflation, and the short-term interest rate. We denote Foreign variables with an asterisk (*), and express all variables, V_t , in logs as $v_t \equiv \ln(V_t)$. To denote the deviation of a variable, V_t , in logs from its steady state, V , we use the notation $\hat{v}_t \equiv \ln(V_t/V)$. Similarly, we denote the deviation of the potential (or frictionless) value of that same variable from its steady state as $\hat{\bar{v}}_t \equiv \ln(V_t^n/V^n)$.

The model of Martínez-García and Wynne (2010) highlights the extent of the international transmission mechanism that arises through trade in goods, while keeping the simplicity and tractability of the work-

¹⁰In a recent work, Eickmeier and Pijnenburg (2013) showed evidence from 24 OECD countries that the common component of changes in unit labor costs has an important effect on inflation. They also considered movements in import price inflation, world interest rate as well as foreign competition and found that these global variables affect inflation.

horse (closed-economy) NK model.

Aggregate demand is described by a pair of open-economy IS equations that links the Home and Foreign output gaps, \hat{x}_t and \hat{x}_t^* , to shifts in consumption demand over time and across countries as captured by domestic and foreign real interest rates, \hat{r}_t and \hat{r}_t^* , in deviations from their respective natural (real) rates $\hat{\bar{r}}_t$ and $\hat{\bar{r}}_t^*$, i.e.,

$$\gamma (1 - 2\zeta) (\mathbb{E}_t [\hat{x}_{t+1}] - \hat{x}_t) \approx ((1 - 2\zeta) + \Gamma) [\hat{r}_t - \hat{\bar{r}}_t] - \Gamma [\hat{r}_t^* - \hat{\bar{r}}_t^*], \quad (1)$$

$$\gamma (1 - 2\zeta) (\mathbb{E}_t [\hat{x}_{t+1}^*] - \hat{x}_t^*) \approx -\Gamma [\hat{r}_t - \hat{\bar{r}}_t] + ((1 - 2\zeta) + \Gamma) [\hat{r}_t^* - \hat{\bar{r}}_t^*], \quad (2)$$

where $\Gamma \equiv \zeta [\sigma\gamma + (\sigma\gamma - 1) (1 - 2\zeta)]$. The real interest rates in the Home and Foreign country are defined by the Fisher equation as $\hat{r}_t \equiv \hat{i}_t - \mathbb{E}_t [\hat{\pi}_{t+1}]$ and $\hat{r}_t^* \equiv \hat{i}_t^* - \mathbb{E}_t [\hat{\pi}_{t+1}^*]$ respectively, \hat{i}_t and \hat{i}_t^* are the Home and Foreign short-term nominal interest rates, and $\hat{\pi}_t$ and $\hat{\pi}_t^*$ are the Home and Foreign inflation rates. The natural real rates that would prevail under flexible prices are denoted $\hat{\bar{r}}_t$ for the Home country and $\hat{\bar{r}}_t^*$ for the Foreign country.

The assumption of price stickiness explains the wedge in the open-economy IS equations between the real interest rate and the natural real rate of interest that captures the distortionary effects of these nominal rigidities on aggregate demand, as shown in Eqs. (1) – (2). However, the Calvo parameter α , which determines the degree of nominal rigidities assumed by the model, does not appear explicitly in the coefficients of these equations. The share of home goods ζ plays a prominent role in the open-economy IS equations as it directly affects the contributions of demand distortions arising in the local and export markets to the dynamic of the output gap of each country.

Aggregate supply is represented with an open-economy Phillips curve relating each country's inflation $\hat{\pi}_t$ and $\hat{\pi}_t^*$, and to domestic and foreign output gaps, \hat{x}_t and \hat{x}_t^* , i.e.,

$$\hat{\pi}_t - \bar{\pi}_t \approx \beta \mathbb{E}_t (\hat{\pi}_{t+1} - \bar{\pi}_{t+1}) + \left(\frac{(1 - \alpha)(1 - \beta\alpha)}{\alpha} \right) [((1 - \zeta)\varphi + \Theta\gamma) \hat{x}_t + (\zeta\varphi + (1 - \Theta)\gamma) \hat{x}_t^*], \quad (3)$$

$$\hat{\pi}_t^* - \bar{\pi}_t^* \approx \beta \mathbb{E}_t (\hat{\pi}_{t+1}^* - \bar{\pi}_{t+1}^*) + \left(\frac{(1 - \alpha)(1 - \beta\alpha)}{\alpha} \right) [(\zeta\varphi + (1 - \Theta)\gamma) \hat{x}_t + ((1 - \zeta)\varphi + \Theta\gamma) \hat{x}_t^*], \quad (4)$$

where $\Theta \equiv (1 - \zeta) \left[\frac{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\zeta)}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\zeta)^2} \right]$. The aggregate supply equation of Martínez-García and Wynne (2010) is augmented with the perceived rates of trend inflation at time t , $\bar{\pi}_t$ and $\bar{\pi}_t^*$ respectively. These equations represent the log-linear approximation to the aggregate dynamics of inflation in a two-country model with Calvo (1983) sticky prices—which flesh out the *global slack hypothesis* relative global (rather than local) slack to local inflation. The variant of the model that we propose here follows Yun (1996), so in periods when firms do not re-optimize their prices they still get to increase them at the perceived trend inflation rate. Departures of aggregate output from potential captured by the respective Home and Foreign output gaps create the gap between inflation and trend inflation modelled in Eqs. (3) – (4).

Monetary policy rule is represented with a reaction function à la Taylor (1993), where the central bank in each country targets domestic inflation and output gap with the domestic short-term nominal interest rate,

$$\hat{i}_t \approx \bar{\pi}_t + [\Psi_\pi (\hat{\pi}_t - \bar{\pi}_t) + \Psi_x \hat{x}_t] + \hat{v}_t, \quad (5)$$

$$\hat{i}_t^* \approx \bar{\pi}_t^* + [\psi_\pi (\hat{\pi}_t^* - \bar{\pi}_t^*) + \psi_x \hat{x}_t^*] + \hat{v}_t^*, \quad (6)$$

where $\tilde{\pi}_t$ and $\tilde{\pi}_t^*$ are the Home and Foreign central bank's target inflation rates, respectively. We assume that the inflation target for each country follows a random walk, i.e.,

$$\tilde{\pi}_t = \tilde{\pi}_{t-1} + \hat{\varepsilon}_t, \quad (7)$$

$$\tilde{\pi}_t^* = \tilde{\pi}_{t-1}^* + \hat{\varepsilon}_t^*, \quad (8)$$

where $\hat{\varepsilon}_t$ and $\hat{\varepsilon}_t^*$ are i.i.d. shocks with zero mean. In our implementation with simulated data, we collapse the target rate to a constant and normalize it to zero to maintain the steady state at zero. However, we maintain the generality of the specification here for illustration purposes.

In this setting, the Home trend inflation $\bar{\pi}_t$ corresponds in equilibrium to the Home central bank's inflation target $\tilde{\pi}_t$. Similarly, the Foreign trend inflation $\bar{\pi}_t$ equals the Foreign central bank's inflation target $\tilde{\pi}_t^*$. To see that, one can interpret the indexation rate $\bar{\pi}_t$ as the Beveridge-Nelson (stochastic) trend of the domestic inflation process,

$$\bar{\pi}_t = \lim_{j \rightarrow \infty} \mathbb{E}_t (\tilde{\pi}_{t+j}). \quad (9)$$

The country's inflation rate $\hat{\pi}_t$ in this model fluctuates around a stochastic trend given by the central bank's inflation target. Hence, since we assume that the target follows a random walk, it must be the case that $\mathbb{E}_t (\tilde{\pi}_{t+j}) = \tilde{\pi}_t$ at any period $j > 0$. In that case, it results from the definition in (9) that $\bar{\pi}_t = \tilde{\pi}_t$ at every point in time. This shows that trend and target inflation must be equal in equilibrium.

Domestic money growth is derived by first-differencing a log-linear *money demand* equation implicitly motivated by a money-in-the-utility-function argument, i.e.,

$$\Delta \hat{m}_t \approx \Delta \hat{c}_t - \eta \Delta \hat{i}_t + \hat{\pi}_t, \quad (10)$$

$$\Delta \hat{m}_t^* \approx \Delta \hat{c}_t^* - \eta \Delta \hat{i}_t^* + \hat{\pi}_t^*, \quad (11)$$

where \hat{m}_t and \hat{m}_t^* are the Home and Foreign stock of money. We also define the natural interest rate as the weighted average of expected domestic and foreign productivity growth,

$$\hat{r}_t \approx \gamma \left(\frac{1+\varphi}{\gamma+\varphi} \right) [(\Theta\Lambda + (1-\Theta)(1-\Lambda)) \mathbb{E}_t [\Delta \hat{a}_{t+1}] + (\Theta(1-\Lambda) + (1-\Theta)\Lambda) \mathbb{E}_t [\Delta \hat{a}_{t+1}^*]], \quad (12)$$

$$\hat{r}_t^* \approx \gamma \left(\frac{1+\varphi}{\gamma+\varphi} \right) [(1-\Theta)\Lambda + \Theta(1-\Lambda)) \mathbb{E}_t [\Delta \hat{a}_{t+1}] + ((1-\Theta)(1-\Lambda) + \Theta\Lambda) \mathbb{E}_t [\Delta \hat{a}_{t+1}^*]], \quad (13)$$

where $\Lambda \equiv 1 + (\sigma\gamma - 1) \left[\frac{\gamma\xi 2(1-\xi)}{\varphi(\sigma\gamma - (\sigma\gamma - 1)(1-2\xi)^2) + \gamma} \right]$. The potential output as the weighted average of Home and Foreign productivity,

$$\hat{y}_t \approx \left(\frac{1+\varphi}{\gamma+\varphi} \right) [\Lambda \hat{a}_t + (1-\Lambda) \hat{a}_t^*], \quad (14)$$

$$\hat{y}_t^* \approx \left(\frac{1+\varphi}{\gamma+\varphi} \right) [(1-\Lambda) \hat{a}_t + \Lambda \hat{a}_t^*], \quad (15)$$

the Home and Foreign output gap as,

$$\hat{x}_t = \hat{y}_t - \hat{y}_t^*, \quad (16)$$

$$\hat{x}_t^* = \hat{y}_t^* - \hat{y}_t, \quad (17)$$

and, finally, the terms of trade and terms of trade gap as

$$\widehat{tot}_t \approx \frac{(\hat{y}_t - \hat{y}_t^*)}{\sigma - (\sigma - \frac{1}{\gamma})(1 - 2\xi)^2} \text{ and } \hat{z}_t \approx \frac{(\hat{x}_t - \hat{x}_t^*)}{\sigma - (\sigma - \frac{1}{\gamma})(1 - 2\xi)^2}, \quad (18)$$

respectively. Note that in this mode the real exchange rate is proportional to the terms of trade, i.e., $\hat{r}s_t = \widehat{tot}_t(2\xi - 1)$.

Finally, the law of motion for productivity shocks and monetary shocks is governed by

$$\begin{pmatrix} \hat{a}_t \\ \hat{a}_t^* \end{pmatrix} \approx \begin{pmatrix} \delta_a & \delta_{a,a^*} \\ \delta_{a,a^*} & \delta_a \end{pmatrix} \begin{pmatrix} \hat{a}_{t-1} \\ \hat{a}_{t-1}^* \end{pmatrix} + \begin{pmatrix} \hat{\varepsilon}_t^a \\ \hat{\varepsilon}_t^{a^*} \end{pmatrix}, \quad (19)$$

$$\begin{pmatrix} \hat{\varepsilon}_t^a \\ \hat{\varepsilon}_t^{a^*} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_a^2 & \rho_{a,a^*} \sigma_a^2 \\ \rho_{a,a^*} \sigma_a^2 & \sigma_a^2 \end{pmatrix} \right), \quad (20)$$

$$\begin{pmatrix} \hat{v}_t \\ \hat{v}_t^* \end{pmatrix} \approx \begin{pmatrix} \delta_v & 0 \\ 0 & \delta_v \end{pmatrix} \begin{pmatrix} \hat{v}_{t-1} \\ \hat{v}_{t-1}^* \end{pmatrix} + \begin{pmatrix} \hat{\varepsilon}_t^v \\ \hat{\varepsilon}_t^{v^*} \end{pmatrix}, \quad (21)$$

$$\begin{pmatrix} \hat{\varepsilon}_t^v \\ \hat{\varepsilon}_t^{v^*} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_v^2 & \rho_{v,v^*} \sigma_v^2 \\ \rho_{v,v^*} \sigma_v^2 & \sigma_v^2 \end{pmatrix} \right). \quad (22)$$

3 Key Implications for Forecasting Inflation

In order to clarify the dynamics of the two-country model in Martínez-García and Wynne (2010), we use the decomposition method advocated by Aoki (1981) and Fukuda (1993) to re-express the core linear rational expectations system that characterizes the log-linearized solution into two separate sub-systems that can be solved separately to characterize the dynamics of the world economy and the differential between the Home and Foreign countries.

The dynamics of the world economy and difference economy are described in Appendix B. Using the definitions for global and differential variables, we move on to the dynamics of country-level inflation to show our key forecasting relationships.

3.1 Dynamics of Country-Level Inflation

We describe the dynamics of the domestic economy for the Home country only, but the approach is analogous to derive the same implications for the Foreign country inflation. We can express the forecast for domestic inflation in terms of the forecasts of global inflation and the inflation differential (their dynamics are presented in Appendix B) with the following transformation,

$$\mathbb{E}_t(\hat{\pi}_{t+j}) = \mathbb{E}_t(\hat{\pi}_{t+j}^W) + \frac{1}{2} \mathbb{E}_t(\hat{\pi}_{t+j}^R).$$

Given the definitions of the slack differential ($\hat{x}_t^R \equiv \hat{x}_t - \hat{x}_t^*$) and the global slack ($\hat{x}_t^W \equiv \frac{1}{2}\hat{x}_t + \frac{1}{2}\hat{x}_t^* = \hat{x}_t - \frac{1}{2}\hat{x}_t^R$) and using the forecasting implications for global inflation and inflation differentials, it follows that,

$$\begin{aligned}\mathbb{E}_t(\hat{\pi}_{t+j}) &= \bar{\pi}_t^W + \frac{1}{2}\bar{\pi}_t^R \\ &= \left(\hat{\pi}_t^W - \frac{\lambda^W}{\mu^W}\hat{x}_t^W\right) + \frac{1}{2}\left(\hat{\pi}_t^R - \frac{\lambda^R}{\mu^R}\hat{x}_t^R\right) \\ &= \hat{\pi}_t - \frac{\lambda^W}{\mu^W}\hat{x}_t^W - \frac{1}{2}\frac{\lambda^R}{\mu^R}\hat{x}_t^R.\end{aligned}$$

Simply re-arranging, we can also express this forecast as follows,

$$\mathbb{E}_t(\hat{\pi}_{t+j} - \hat{\pi}_t) = -\frac{\lambda^W}{\mu^W}\hat{x}_t^W - \frac{1}{2}\frac{\lambda^R}{\mu^R}\hat{x}_t^R. \quad (23)$$

Note that forecasting future inflation using the foreign and domestic output gaps alone would not be accurate since domestic inflation potentially has a stochastic trend while foreign and domestic slack are stationary; one needs to include among the regressors some variable with a similar stochastic trend to that of domestic inflation. But this need not be money growth or the terms of trade; current inflation itself has the same stochastic trend, so including it to forecast future inflation takes care of the trend component without the need to include any other regressors to attempt to track the stochastic trend. However, based on our previous analysis we can use the real world money demand gap to replace the hard-to-measure global slack and forecast domestic inflation with domestic slack and the global real money balances gap. In turn, we can use the terms of trade or real exchange rate gap instead of the slack differential and equally forecasting domestic inflation with the help of domestic slack and the terms of trade gap (or the real exchange rate gap). The key insight is summarized in the following proposition,

Proposition 1 *No variables other than domestic and foreign slack should help improve the forecast of changes in domestic inflation. The forecasting relationship for domestic inflation implied by the workhorse Open-Economy NK model of Martínez-García and Wynne (2010) can be expressed as,*

$$\mathbb{E}_t(\hat{\pi}_{t+j}) = \hat{\pi}_t - \frac{1}{2}\left(\frac{\lambda^W}{\mu^W} + \frac{\lambda^R}{\mu^R}\right)\hat{x}_t - \frac{1}{2}\left(\frac{\lambda^W}{\mu^W} - \frac{\lambda^R}{\mu^R}\right)\hat{x}_t^* \quad (24)$$

$$= \hat{\pi}_t - \frac{\lambda^W}{\mu^W}\hat{x}_t + \left(\frac{\lambda^W}{\mu^W} - \frac{\lambda^R}{\mu^R}\right)\frac{1}{2}\hat{x}_t^R \quad (25)$$

$$= \hat{\pi}_t - \frac{\lambda^R}{\mu^R}\hat{x}_t - \left(\frac{\lambda^W}{\mu^W} - \frac{\lambda^R}{\mu^R}\right)\hat{x}_t^W, \quad (26)$$

where global slack and differential slack can be proxied by the global real money demand gap and the terms of trade

gap (or real exchange rate gap) as follows,

$$\begin{aligned}\hat{x}_t^W &\equiv \frac{1}{2}\hat{x}_t + \frac{1}{2}\hat{x}_t^* \approx \frac{1}{\chi}\hat{m}_t^{g,W}, \\ \hat{x}_t^R &\equiv \hat{x}_t - \hat{x}_t^* \approx \frac{1}{\kappa^{tot}}\left(\widehat{tot}_t - \widehat{tot}_t^*\right) \approx \frac{1}{\kappa^{rs}}\left(\widehat{rs}_t - \widehat{rs}_t^*\right),\end{aligned}$$

where $\chi \equiv \left(1 - \eta \left(-\psi \pi \frac{\lambda^W}{\mu^W} + \psi_x\right)\right)$, $\kappa^{tot} \equiv \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(2\xi - 1)^2}\right]$ and $\kappa^{rs} \equiv (2\xi - 1) \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(2\xi - 1)^2}\right]$.

Proof. See proofs for Propositions 2 and 3 in Appendix B. ■

More generally, using the fact that $\hat{\pi}_{t+h|t}^h \approx \frac{400}{h} \sum_{j=1}^h \hat{\pi}_{t+j}$, we can write the actual forecast h -periods ahead for domestic inflation as follows,

$$\begin{aligned}\mathbb{E}_t \left(\hat{\pi}_{t+h|t}^h \right) &= \frac{400}{h} \sum_{j=1}^h \mathbb{E}_t \left(\hat{\pi}_{t+j} \right) = \frac{400}{h} \sum_{j=1}^h \mathbb{E}_t \left(\bar{\pi}_{t+j} \right) \\ &= 400 \left(\hat{\pi}_t - \frac{1}{2} \left(\frac{\lambda^W}{\mu^W} + \frac{\lambda^R}{\mu^R} \right) \hat{x}_t - \frac{1}{2} \left(\frac{\lambda^W}{\mu^W} - \frac{\lambda^R}{\mu^R} \right) \hat{x}_t^* \right)\end{aligned}\quad (27)$$

$$= 400 \left(\hat{\pi}_t - \frac{\lambda^W}{\mu^W} \hat{x}_t + \left(\frac{\lambda^W}{\mu^W} - \frac{\lambda^R}{\mu^R} \right) \frac{1}{2} \hat{x}_t^R \right) \quad (28)$$

$$= 400 \left(\hat{\pi}_t - \frac{\lambda^R}{\mu^R} \hat{x}_t - \left(\frac{\lambda^W}{\mu^W} - \frac{\lambda^R}{\mu^R} \right) \hat{x}_t^W \right). \quad (29)$$

This implies that no other variable should improve our forecast of domestic inflation if domestic slack and foreign slack are included in the forecasting model. This feature is similarly noted in Woodford (2008) for a closed-economy, and we use it as our key identifying restriction in order to construct a reduced-form specification (an autoregressive distributed lag model) for forecasting inflation that is consistent with the open-economy Phillips curve of Martínez-García and Wynne (2010).

3.2 Benchmark Calibration

The model parameters are summarized in Table 1a below.¹¹ We calibrate the two countries to the U.S. and Europe in line with Chari et al. (2002).¹² Our calibration strategy is two-fold. For parameters that are of particular interest for our study, we use the (median) values of historical estimates of these parameters. For the remaining, we refer to the literature. We set $\beta = 0.99$, $\gamma = \varphi = 5$, and $\alpha = 0.75$, as in Chari et al. (2002). This is also similar to the closed-economy model in Neiss and Nelson (2003) and Neiss and Nelson (2005). We assume that countries are equal in population, $n = 0.5$, and the allocation of home and foreign goods in the consumption basket of each country is symmetric and given by the common parameter ξ . We set the interest semi-elasticity of money demand $\eta = 4$, as described in Galí (2008). We assume that the cross-country spillover of the productivity shocks, δ_{aa^*} , is zero. The correlations of domestic and foreign productivity innovations ρ_{a,a^*} are set as zero as well, while the correlation of domestic and foreign monetary

¹¹An alternative could be to estimate the model, as in Benati and Surico (2008). One possible argument in favor of calibration, however, is that the model may be too simplified. Hence one may be concerned that estimating it would lead to a misspecification bias and, therefore, would complicate the interpretation of parameter estimates and our subsequent experiments even more.

¹²Chari et al. (2002) considers an aggregate of France, Germany, Italy, and the U.K. for 1973:Q1-1994:Q4.

innovations, ρ_{v,v^*} , is set as 0.5, following Chari et al. (2002).¹³ We assume further that the monetary and productivity innovations are uncorrelated with each other.

For the remaining parameters which are key for our analysis, we conduct rolling window estimations. We use the median of these estimates in the benchmark calibration for 2004:Q1-2008:Q2. Our calibration strategy is summarized in Table 1b. Appendix C presents further details of the estimation and the results. Our findings suggest the presence of structural shifts on a number of key parameters. This motivates us to explore three channels on the forecasting performance of standard forecasting variables from open-economy Phillips curves for U.S. inflation. For this reason, we investigate how good luck, changes in the conduct of monetary policy, and the effects of globalization play a role in determining the observed forecasting performances.

In order to obtain historical estimates, we conduct 20-year rolling window regressions on the processes underlying the behavior of the Solow residual, the Taylor rule and the corresponding monetary policy shock, as well as the import shares and the price elasticity of trade. These estimates are solely based on U.S. data and abstract from some other aspects that are essential in the model for lack of comparable data—like the extent of international linkages, for instance, in the case of estimating spillovers and covariances for the Solow residual. However, our forecast results are also robust to parameter values suggested by the literature for cross-country spillovers and covariances. The choice of the starting and ending dates of these periods are determined by the observed structural breaks as well as limitations on data.¹⁴ Hence, we suggest that there are two distinct periods to consider, 1973:Q4-1983:Q4 and 2004:Q1-2008:Q2.¹⁵ Therefore, we put the emphasis in understanding, broadly, the shifts that occurred between the late 1960s and the Great Moderation period prior to the 2008 recession.

While the first period is characterized by a very large probability that the policy rule violated the Taylor principle, we leave the analysis of the forecasting performance during the period of indeterminacy prior to that for future research.¹⁶ Accordingly, we assume that the Taylor rule is noninertial and the policy rule is identical in both countries. We use the U.S. data and obtain the two policy rule parameter estimates $\Psi_\pi = 1.76$ and $\Psi_\chi = 0.74$. The persistence and volatility parameters of the AR(1) monetary shock process are $\delta_v = 0$ and $\sigma_v = 1.31$, respectively. For the productivity shock, the persistence and volatility parameters are $\delta_a = 0.81$ and $\sigma_a = 0.69$, respectively. The parameters of the shock processes are the same for the two countries. The European share of goods in the U.S. consumption basket, ξ is chosen as 0.08. The elasticity of substitution between U.S. and European goods, σ , is set at 0.97.

¹³We also tried other values for these parameters, where we set $\delta_{aa^*} = 0.025$ and $\rho_{a,a^*} = 0.25$, as suggested by Heathcote and Perri (2002), and found the results are robust to this alternative parameterization.

¹⁴We mark important dates for monetary policy changes as follows. The period for Volcker's Policy of Targeting Monetary Aggregates is October 1979 - October 1982. Greenspan took office in August of 1987. Also note that the 2001 recession started in the first quarter; and that the 2008 recession started officially in the fourth quarter of 2007 in the U.S. but it did not become severe until the second or third quarter of 2008.

¹⁵Hence our historical estimates for the model's parameters use data for the two periods: 1953:Q4-1983:Q4 and 1984:Q1-2008:Q2.

¹⁶Although the Taylor principle is neither necessary nor sufficient for determinacy in this open-economy model, it is a very good short-hand approximation for the region of the parameter space where the model would be determinate as shown in Martínez-García and Wynne (2010).

Table 1a: Model parameters	
Structural parameters	
Intertemporal discount factor	$0 < \beta < 1$
Inverse of the intertemporal elasticity of substitution	$\gamma > 0$
Inverse of the Frisch elasticity of labor supply	$\varphi > 0$
Interest semi-elasticity of money demand	$\eta > 0$
Elasticity of substitution across varieties within a country	$\theta > 1$
Elasticity of substitution between Home and Foreign bundles	$\sigma > 0$
Share of Home goods in the Home basket	$0 < \xi < 1$
Home population size, Mass of Home varieties	$0 < n < 1$
Foreign population size, Mass of Foreign varieties	$0 < 1 - n < 1$
Calvo (1983) price stickiness parameter	$0 < \alpha < 1$
Monetary policy parameters	
Sensitivity to deviations from the inflation target	$\Psi_\pi > 1$
Sensitivity to deviations from the potential output target	$\Psi_x > 0$
Shock parameters	
Persistence of the productivity shock	$-1 < \delta_a < 1$
Cross-country spillover of the productivity shocks	$-1 < \delta_{aa^*} < 1$
Volatility of the productivity shock	$\sigma_a > 0$
Correl. between Home and Foreign productivity innovations	$-1 < \rho_{a,a^*} < 1$
Persistence of the monetary policy shock	$-1 < \delta_v < 1$
Cross-country spillover of the monetary policy shock	$-1 < \delta_{v,v^*} < 1$
Volatility of the monetary policy shock	$\sigma_v > 0$
Correl. between Home and Foreign monetary innovations	$-1 < \rho_{v,v^*} < 1$

While we calibrate the model to the median values of the parameter estimates for the (benchmark) 2004:Q1-2008:Q2 period, we also consider other plausible values that lie within the upperbounds and lowerbounds of these parameter values using the 95% confidence bands of our estimates in the benchmark period as well as in the (counterfactual) 1973:Q4-1983:Q4 period.

To produce data, we run a Monte Carlo simulation of the model with 100 trials and with a subsample of 160 periods for each trial. We split the 160 periods equally between the estimation sample and the pseudo out-of-sample forecast sample to be consistent with our empirical analysis.

Using the simulated data, we forecast inflation using two-variable and three-variable recursive forecasts. We run 1–, 4–, and 12–quarters ahead inflation forecasts where we test the various forms of the reduced-form Phillips curve. In particular, we calculate the (relative) mean square forecast errors (MSFEs) at a grid of points that spans the space for selected parameters, while keeping other parameters at their benchmark values. We conduct the grid search symmetrically for the two countries. In these 100 trials, we evaluate forecasting performance based on the median (relative) MSFE, the median p-value of the hypothesis that the relative MSFE is greater than or equal to 1, and the fraction of statistically significant trials with p-values less than or equal to 5 percent. The details of our forecast methodology are described in the next section. The experiments with simulated data conducted here can be grouped under three competing hypotheses:

i) Good luck, ii) Monetary policy, and iii) Openness.

i) Our *good luck* experiment focuses on how forecasting performance of the regressors listed above is altered when the parameters of the shock processes, specifically the volatility of shocks, σ_v and σ_a take on a range of values.¹⁷ We perform this exercise symmetrically for both countries: σ_v varies within $[1.06, 3.16]$ and σ_a takes values within $[0.59, 1.83]$. Notice that these intervals are calculated based on the highest and lowest estimates obtained in the two periods (This applies to all three experiments.)

ii) Our *monetary policy* experiments pay attention to forecasting performance under changes in the monetary policy parameters Ψ_π and Ψ_x . We discard estimates that lie in the indeterminacy region. Hence, for Ψ_π , we try values of grid points in the interval $(1, 2.42]$ and Ψ_x , in the interval $(0, 0.99]$.

iii) The final experiment, *trade openness*, involves a grid search over the parameter for the share of foreign goods in the domestic basket, ζ and the elasticity of substitution between Home and Foreign bundles, σ . For ζ we try the values in the intervals $(0, 0.4]$. Under the case that ζ is close to 0, the economy is almost closed. Here, we use a range that is broader than the interval defined by our estimates, acknowledging the fact that openness can be more broadly defined in reality and with our estimates the experiments might be underestimating the effects of openness on forecasting. For σ , we try values within the range $[0.16, 1.81]$. Note that $\sigma = 1$ implies the consumption aggregator is Cobb-Douglas type.

Table 1b: Calibration			
		1973:Q4-1983:Q4	2004:Q1-2008:Q2
		Counterfactual period	Benchmark period
Good luck	σ_a	$[1.03, 1.83]$	$[0.59, 0.97]$
	σ_v	$[1.06, 3.16]$	$[1.10, 1.88]$
Monetary policy	Ψ_π	$(1, 1.81]$	$[1.30, 2.42]$
	Ψ_x	$(0, 0.72)$	$(0, 0.99]$
Openness	σ	$[0.16, 1.37]$	$[0.16, 1.81]$
	ζ	$[0.025, 0.0305]$	$[0.048, 0.096]$
Benchmark values (period averages)	σ_a	1.30	0.69
	σ_v	1.41	1.31
	Ψ_π	1.02	1.76
	Ψ_x	0.37	0.74
	σ	0.81	0.97
	ζ	0.03	0.08
	δ_a	0.93	0.81
	δ_v	0.00	0.00
	δ_{aa^*}	0.00	0.00
	ρ_{aa^*}	0.00	0.00
	ρ_{vv^*}	0.50	0.50

¹⁷In our current terminology, ‘good luck’ refers to the possibility of a period of low volatility. Good luck might also be the result of an unusual draw of benign shocks from the right-tail of the distribution. Rather, we interpret good luck as the shift in the distribution of shocks that reduces exogenous volatility.

3.3 Simulation

3.3.1 Forecast Models

Following Stock and Watson (2003a), we refer to models with explanatory variables as economic models and we assess with simulated data to what extent these economic models represent an improvement over the univariate model of forecasting inflation. Note also that the forecasting models introduced in this section are the same as the models used later in the empirical section of the paper.

While Atkeson and Ohanian (2001) cast doubt on the predictive ability of Phillips curve-based forecasts, Stock and Watson (1999a), Stock and Watson (1999b) and Stock and Watson (2008) provide some empirical evidence in favor of the Phillips curve as a forecasting tool, suggesting that inflation forecasts generated with Phillips curve-based regressors generally are more accurate than forecasts based on other macroeconomic variables (including interest rates, money, and commodity prices). We expand on this literature going beyond the conventional closed-economy Phillips curve framework and show with simulated data (as well as in theory) that measures of global economic activity can yield more accurate forecasts than the naïve model and the closed-economy Phillips curve.

In order to test the global slack hypothesis in theory, we use global slack and variables that have information content for global slack as shown in our theoretical results under the Martínez-García and Wynne (2010) framework.

First, we consider the reduced-form representation of a closed-economy NK Phillips curve relating inflation to domestic real economic activity,

$$\hat{\pi}_{t+h|t}^h = a_1 + \lambda_{11}(L) \hat{\pi}_t + \lambda_{12}(L) \hat{x}_t + \hat{\epsilon}_{1,t+h}. \quad (\text{Model 1})$$

By denoting the quarterly forecast horizon as h , it is possible to forecast h -quarters ahead inflation, $\hat{\pi}_{t+h|t}^h$, with the distributed lag of earlier inflation rates, $\hat{\pi}_t$, to proxy for expected future inflation, and the distributed lag of the domestic economic measure, \hat{x}_t (i.e. domestic slack, or money growth as a proxy for slack). We define h -quarter ahead (annualized) inflation $\hat{\pi}_{t+h|t}^h = \frac{400}{h} \times [\ln(P_{t+h}/P_t)]$, and forecast inflation for horizons ranging from 1-quarter-ahead to 12-quarters-ahead.

Following on Woodford (2008)'s rationale, there should be no other variable than global slack to forecast changes in domestic inflation. The lags of inflation help us take into account the unmodeled stochastic trend without the need to use an additional variable that has a similar trend. The number of lags for each variable is selected based on the Schwarz information criterion (SIC). To keep the model parsimonious, and since the frequency of the variables is defined as quarterly, the maximum possible lags allowed for each variable is set as four. Note that this specification is also similar to the models employed in the empirical literature, so it has the advantage to facilitate comparison.

Following from Proposition 1, in order to test the performance of global slack (as described by two variables, domestic and foreign slack) and its proxies (domestic slack and terms of trade gap, or alternatively, domestic and foreign money growth), we consider the open-economy New Keynesian Phillips curve in the following reduced-form,

$$\hat{\pi}_{t+h|t}^h = a_2 + \lambda_{21}(L) \hat{\pi}_t + \lambda_{22}(L) \hat{x}_t + \lambda_{23}(L) \hat{z}_t + \hat{\epsilon}_{2,t+h}. \quad (\text{Model 2})$$

Under this specification, \hat{x}_t and \hat{z}_t denote, respectively, either (i) domestic and foreign slack, (ii) domestic slack and terms of trade gap, or (iii) domestic and foreign money growth.

We underline some important points here. First, we use first-differenced terms of trade as a terms of trade gap measure, rather than using the theoretically-consistent variable because the theoretical terms of trade gap cannot be obtained in practice. The first-differenced terms of trade can still proxy the true terms of trade gap in theory, and this is supported by our quantitative experiments. Second, there is no need to test the performance of the real exchange rate gap as it would yield the same results as the terms of trade gap.¹⁸ Third, the model does not differentiate between terms of trade and terms of trade ex. oil, so we leave the analyses of these different variables to the empirical section. And finally, we use world (nominal) money growth to proxy for the world real money gap which is shown to be proportional to global slack. One could argue that real world money growth should be used as a proxy for the world real money gap instead, but then real world money growth could be expressed in terms of the difference between nominal world money growth and world inflation. Hence, nominal world money growth should be a sufficient proxy for global slack and has information content for inflation.

Finally, in order to compare the performance of these models against a benchmark, we introduce the ‘restricted’ model specified as a univariate autoregressive (AR) process

$$\hat{\pi}_{t+h|t}^h = a_3 + \lambda_3(L) \hat{\pi}_t + \hat{\epsilon}_{3,t+h}. \quad (\text{Model 3})$$

To be consistent with the theoretical framework, we consider inflation series in levels. We make a similar argument as in Woodford (2008). In all three models, current inflation itself has the same stochastic trend, so including it to forecast future inflation takes care of the trend component without the need to include any other regressors to attempt to track the stochastic trend.

3.3.2 Forecast Scheme

We perform forecasts based on the pseudo out-of-sample forecasting method and particularly focus on recursive samples. Therefore, at any given date t , we forecast inflation at date $t + h$ using all available data up to date t . The models are estimated by OLS. We assess the multi-step pseudo out-of-sample forecasting performance of a model that incorporates variables commonly thought as contemporaneous or leading indicators of inflation relative to the forecast of a univariate AR process.

Our forecast evaluation metric, the relative MSFE, is the ratio of the MSFE of the economic model (Model 1 or Model 2) relative to that of the benchmark AR model (Model 3). Let T_0 denote the starting date of the data series and T_1 denote the end. The estimation sample starts at T_0 and ends in t_0 . We start by using all data up to date t_0 to forecast inflation at date $t_0 + h$. By adding data to the estimation sample, we keep estimating the parameters of the model of interest. The h -step recursive forecast continues until period $T_1 - h$ with a total of $T_1 - h - t_0 + 1$ steps. For a given model j , this procedure yields a sequence of forecast errors which helps us construct the MSFE of the model at horizon h and from date t_0 to $T_1 - h$,

$$MSFE_j(h) = \frac{1}{T_1 - h - t_0 + 1} \sum_{t=t_0}^{T_1-h} \hat{\epsilon}_{j,t+h}^2$$

¹⁸Obviously, there is a pronounced discrepancy between terms of trade and REER results in the empirical section we will present later, since the two variables are not perfectly correlated in the actual data.

where $\hat{\epsilon}_{j,t+h}$ is the estimated forecast error for model j at date $t + h$.

3.3.3 Inference and Samples

Inference is based on the F-statistics against critical values based on a bootstrap algorithm described in Clark and McCracken (2006).¹⁹ We extend the methodology of Clark and McCracken (2006) for three-variable forecasts. (The details of the bootstrap algorithm with three variables are described in the Appendix E.) This procedure involves resampling from the residuals of vector autoregressive (VAR) equations. In order to test the predictive ability of a single variable forecast as in Model 1, we define an equation for inflation (as governed by the restricted model 3) and an equation for the predicting variable, where the lag length for the predicting variable and inflation are separately determined based on the SIC. The equations of the data generation process (DGP) are estimated by OLS with a number of bootstrap iterations equal to 100. (In the empirical part, the number of iterations are set at 5000.)

We then have a one-sided test with the null hypothesis that an economic model (Model 1 or Model 2) does not yield more accurate forecasts than the AR process (Model 3), i.e. $MSFE_{AR} \leq MSFE_{EM}$, against the alternative that $MSFE_{AR} > MSFE_{EM}$. Throughout the paper, we report the MSFE of the benchmark model and the relative MSFEs of a particular economic model 1 or 2 and the benchmark model 3. The null hypothesis is expressed as ‘the relative MSFE is greater than or equal to 1’. We report the p-values of the F-test at 1% and 5%. (Up to 10% in tables F1-F3).

3.4 Results and Interpretation

In light of our quantitative analyses, we show that an open-economy Phillips curve helps forecast U.S. inflation and moreover, given a plausible calibration of the model, global slack and other global economic variables seem to have gained value in forecasting inflation in the benchmark period relative to the counterfactual period. Hence these variables appear to be good forecasting variables and most importantly, they can be used as proxies for global slack. Figures E1-9 in Appendix E illustrate the simulated forecast results regarding the good luck, monetary policy and openness experiments, respectively. The figures report results under all historically plausible values for the parameters of interest, while we also highlight the median values to facilitate a comparison between Period 2 (benchmark) and Period 1 (counterfactual). The median parameter value for period 1 is indicated with the marker ‘+’ and period 2 with ‘x’.

- The key results from the good luck experiments (in Figures E1-3) are summarized as follows. All Phillips curve-based forecasts (constructed with domestic or domestic and foreign variables), gained accuracy relative to the naïve forecasts during the Great Moderation period compared to the pre-Great Moderation period due to the declines in the volatility of productivity shocks, σ_a . However, a decline in monetary policy shocks, σ_v , all else equal, seems to reduce forecast accuracy in general, although this effect is offset by the improvement due to the lower volatility of the productivity shocks.
- The monetary policy experiments (in Figures E4-6) suggest that it is the parameter on the deviation from the inflation target, Ψ_π , rather than the output gap, Ψ_x , that matters for the changes in forecast

¹⁹The construction of F-statistics as well as t-statistics are described in Clark and McCracken (2001a), Clark and McCracken (2001b), and Clark and McCracken (2002). Inference can also be based on t-statistics, however, as suggested by these authors, F-type tests are more powerful than the corresponding t-type tests, and therefore we focus on F-statistics only.

accuracy given the plausible ranges for the parameter values that we estimated. All else held constant, a higher anti-inflationary bias improves forecasts for both closed and open-economy Phillips curves and for short horizons.

- Our results from the *openness* experiment are shown in Figures E7-9. Increases in both the input shares, ξ , and the elasticity of substitution between exports and imports, σ , only improve the performance relative to naïve forecasts of open-economy Phillips curve-based forecasts.
- In Table 1c below, we show a summary of the simulation results.

Table 1c: Predictive performances of variables				
		Domestic & foreign slack	Domestic slack & ToT (FD)	Domestic & foreign money growth
Good luck	σ_v			
	σ_a	✓	✓	✓
Monetary policy	Ψ_π	✓	✓	✓
	Ψ_x			
Openness	ξ	✓	✓	✓
	σ	✓	✓	✓

Note: This table reports whether changes in a given parameter have a statistically significant impact on the predictive ability of a variable (at 5% significance level or below) in at least 50% of the trials of the experiment.

4 Empirical Analysis

Having established the main findings from the simulated forecasts, we aim to test these results empirically. We perform these forecasting exercises here to reevaluate the role of some of these measures in order to provide a comparison with our main forecasting strategy and to also make an extensive robustness analysis of the existing work. We evaluate the predictive ability of open-economy Phillips curves using global output gap measures, as well as variables that can be shown to be proportional to global slack in theory: (i) global money growth, (ii) global credit growth, (iii) domestic slack and the terms of trade gap, (iv) domestic slack and the real exchange rate gap.

The issue of how to measure the output gap—both domestic and foreign—has been known as a major challenge. For purely statistical approaches which in most cases derive potential output using actual (real) output series through a filtering technique (most commonly the HP filter), the choice of the filter is usually an arbitrary decision. In addition, applying these techniques is known to create end-point problems. For structural estimates of the output gap, relying on a production function (such as Cobb-Douglas) and quantifying the total factor productivity, the capital stock or labor employed tend to pose measurement problems, as well. (Gerlach (2011)).

Measuring the foreign output gap, however, is an even more challenging task since for the emerging market economies that are believed to potentially affect U.S. inflation, the data series to measure unemployment rates or capacity utilization in manufacturing are usually either too short or they are not available.²⁰ Furthermore, there is no clear idea of how the dynamics of the foreign output gap affect domestic inflation. Therefore, estimating the open-economy Phillips curve based on the combination of domestic and foreign slack as a measure of the global slack is even more difficult.

This problem has been addressed also in an earlier work by Martínez-García and Wynne (2010) where they replace global slack in reduced form with a combination of domestic slack and the terms of trade gap. In theory, the terms of trade gap in addition to the domestic output gap has information content for forecasting domestic inflation, as formulated in the open-economy Phillips curve. Since the terms of trade gap and real exchange rate are proportional in theory, one can replace the terms of trade measures with real exchange rate measures. We apply one-sided filters (first-differencing and the HP filter) on the terms of trade and real effective exchange rate series to obtain the terms of trade and real exchange rate gap measures.

The long-run relationship between the growth rate of monetary aggregates and the rate of inflation is explained in the literature by the quantity theory of money, and these measures of money growth are suggested to have information content for inflation forecasting (D’Agostino and Surico (2009)). However, we take a different view here on the predictive ability of money and suggest that money has information content because it is related to the output gap under a standard NK model.²¹ A similar argument could be extended to credit growth, as well—even though it is not modelled in a standard open-economy NK framework of Martínez-García and Wynne (2010). Under this framework, real world money gap is shown to be proportional to the global slack. In order to proxy for the real world money gap, we use global money growth measures (and global credit growth measures, similarly.)

4.1 Data

Figures F1-2 in Appendix F plot the series employed throughout the empirical section. The U.S. inflation rate is calculated as annualized log-differences of quarterly series of six price indexes in our benchmark and rolling window forecasts: consumer price index (CPI), core CPI (CPI ex. food and energy), personal consumption expenditure deflator (PCE), core PCE (PCE ex. food and energy), GDP deflator and producer price index (PPI). In addition, we consider two short inflation series: The Atlanta Fed’s sticky price and sticky price ex. shelter which are used only in the benchmark analysis.

We perform inflation forecasts using a wide range of domestic and global slack measures. Our domestic measures consist of: CBO U.S. slack, FRBD U.S. slack, OECD U.S. slack, IMF U.S. slack and HP-filtered U.S. real GDP. For global slack measures, we use: FRBD G7, FRBD G28, OECD G7, OECD Total and IMF Advanced series. All series are available quarterly, except for the IMF measures of domestic and global slack, which are available at annual frequency. Therefore, we disaggregate these annual series into quarterly frequency using the quadratic match average method.

²⁰Stock and Watson (1999b) evaluate the conventional Phillips curve-based forecast with unemployment and report that it can be improved with broader measures of economic activity. Lack of consistent data on unemployment across countries forces us to rely primarily on related measures of economic activity and slack. Globally, since data availability does not permit us to consider the role of unemployment in our analysis.

²¹In particular, global real money growth is shown to be proportional to global output gap in the standard open-economy New Keynesian model studied in this paper.

The terms of trade series is calculated as the ratio of the U.S. export price index of goods and services to the U.S. import price of goods and services. For terms of trade ex. oil, however, we use the price indexes for exported goods and nonpetroleum imported goods. We use 1-sided filters (first differences, HP filter) in order to obtain a measure of the terms of trade gap. We use the U.S. trade-weighted real effective exchange rate series and apply the same filters to obtain the real exchange rate gap measures.

We define global money growth as the average of the percentage growth rates of monetary aggregates in the G7 countries. While we pick the series for monetary aggregates that are most similar in definition, we are constrained by quarterly data availability for Canada, France, Germany, Italy and Japan especially in the 1960s and 1970s. Since we would like to extend the robustness analysis of forecasting experiments to a large estimation sample, we make our primary decision on selection based on data availability.

We construct a measure of global credit growth, by calculating the G7 average of credit growth rates. We consider quarterly, long series on credit to non-financial sectors in G7 countries. We particularly focus on credit from all sectors to the private sector using the existing BIS data. (A more detailed explanation is available in Appendix F.)

4.2 Forecast Models

Our forecast strategy for the empirical analysis is the same as in the simulated forecasts. In this section however, we only consider bivariate forecasts and a univariate AR process for a benchmark specification, as described by Model 1 and Model 3, respectively. The bivariate economic models are used in order to evaluate the forecast accuracy of individual measures of domestic slack and global slack, domestic and global liquidity growth²², domestic and global credit growth, and terms of trade and real effective exchange rate gap measures.²³

In our benchmark experiments, the estimation sample begins in 1980:Q1 and ends in 1991:Q4 and the pseudo out-of-sample forecasting period begins in 1992:Q1 and 2011:Q4 leaving us with an estimation sample of 48 quarters and the pseudo out-of-sample forecasting sample of 80 quarters.²⁴ The results are reported in Tables F1-3.

In addition to our benchmark forecasting experiment, we conduct a series of other experiments going back in time to the extent that the series are available in order to make a robustness analysis. More specifically, starting with the initial observation in the sample, we shift the estimation and forecast samples backward by one quarter and obtain the relative MSFEs of the forecasts for each ‘rolling window’.²⁵ Each

²²D’Agostino and Surico (2009) evaluate the forecasting performance of the average growth rate of broad money in G7 economies and find that the results are significantly more accurate compared to forecasts with U.S. money growth.

²³Canova (2007) evaluated the performance of nominal and real money growth across G7 economies for the 1996:Q1-2000:Q4 period (or the subperiods) and found these results are comparable to Phillips curve based forecasts.

²⁴Our selection of the size of the estimation and pseudo out-of-sample forecasting samples in the benchmark experiments follow that of D’Agostino and Surico (2009) which enables us to compare the measures used to forecast inflation with their findings. In our robustness analyses, we make a symmetric allocation of the observations for the two samples.

²⁵While we adopt the bootstrap algorithm of Clark and McCracken (2006) for empirical inference with recursive forecasts we did also consider the implementation of the fluctuations test of Giacomini and Rossi (2010) using Giacomini and White (2006) test statistic (Giacomini and Rossi (2010) refer to this as the GW test). This test statistic is also equivalent to Diebold and Mariano (1995) and West (1996) test statistics. Clark and McCracken (2013) note that the Diebold-Mariano-West framework is not supposed to be valid in general for the case of nested models that we have considered in this paper, although it may still work in finite samples. However, Giacomini and Rossi (2010) show in Monte Carlo experiments that the full sample GW test seems to have very low power—so conditional on the null hypothesis of equal forecast accuracy being false, the probability of rejecting such a null is very low. That is essentially what our implementation of the GW test, whose results are available upon request, would suggest. In finite samples with the data we have, we generally find that the GW test imposes a threshold to detect differences in forecasting performance harder to cross than the test of

window spans 80 quarters of an estimation sample and 80 quarters of a forecasting sample. The results are reported in Figures F3-14.

The regression equation for the open-economy Phillips curves that use the domestic output gap and terms of trade (or REER) can be described as in Model 2. In an initial assessment, using Model 1, we found mixed results from forecasting inflation with the domestic output gap and terms of trade gap (measured as first-differenced terms of trade),²⁶ and obtained stronger results with the terms of trade gap or REER gap alone (Model 1). In this regard, weak empirical evidence on some measures of global slack does not necessarily invalidate the global slack hypothesis. Our results indicate that the measures available may be subject to well-known measurement errors.²⁷

We also note that some of the global slack variables that we use come from international organizations such as the IMF and OECD that use their own aggregation methods. The Dallas Fed measures are based on constant PPP-adjusted GDP weights. For our global money supply growth measure, we use a simple arithmetic average following D'Agostino and Surico (2009). For global credit growth, we follow the same aggregation technique. We recognize that there are different aggregation schemes for different variables. However, what we are doing is using indicators most of which have already been introduced in the literature or indicators that have become standard for use in applied work. Even if we adopted a common aggregation scheme, for instance, we would not be able to decompose most of the global slack measures into domestic and foreign components. Therefore, we just accept that differences in aggregation may be another contributing factor explaining some of the weaknesses of the existing empirical evidence. Hence, in the empirical part, we only focus on bivariate forecasts and use Model 1 to test the predictive ability of all Phillips curve forecasts in the empirical section.

4.3 Empirical Findings

The results of the pseudo out-of-sample forecast with one variable over the benchmark sample are reported in Tables F1-3. Our findings can be listed as follows:

1. Based on the one-variable forecast results, it is not possible to say that global slack measures outperform the domestic slack measures. In general, both measures almost equally yield more accurate predictions compared to an AR process when the inflation measure is either core CPI or core PCE. For other measures of headline inflation, however, we conclude that the AR process of inflation performs better.
2. Global money growth (measured as G7 average) exhibits a better forecasting performance relative to U.S. money growth, at all horizons for CPI, core CPI and PCE deflator. Both variables have a significantly poor performance compared to the AR process in all other inflation measures. Under the forecasts of CPI and PCE inflation, G7 money growth also does better compared to domestic or global slack measures. However, this is not true for the other measures of inflation.²⁸

Clark and McCracken (2006) does.

²⁶Similar results can be confirmed when terms of trade is replaced by the real effective exchange rate (first-differenced).

²⁷Clark and McCracken (2006) suggest that in asymptotically nested models, there is a trade-off to adding regressors to the forecast model: the fit may improve with the additional regressor while the estimation error variance increases. This may be another concern for multivariate forecasts, but we mainly focus on that measurement problems of regressors, particularly the slack measures, and the contribution of the additional variables to palliate these problems.

²⁸While real money growth measure is the theoretically relevant one, the nominal money measures help forecast inflation more

3. Forecasting performance of terms of trade is comparable to that of domestic and global slack measures. Terms of trade ex. oil has no significant improvement over the AR specification across any of the inflation measures and at any horizon.

We also perform rolling window experiments for three groups of variables: a domestic slack measure vs. two global slack measures; terms of trade vs. terms of trade ex. oil; domestic vs. global money supply growth; and domestic vs. global credit growth. Among several alternatives, we choose the CBO measure as the domestic slack variable and 'OECD Total' and 'FRBD G7' as our global slack measures. Our selection of the two measures is based mainly on the length of the series and the relatively better performance compared to other slack measures at hand. In Figures F3-14 in Appendix F, we show how the forecasting performances of these pairs of variables evolve over time. In these figures, several interesting points emerge:

1. Terms of trade and terms of trade ex. oil appear to have some value to improve the forecasting performance starting in the mid-1980s (Figures F3-6). There is no regular pattern as to whether terms of trade or terms of trade ex. oil performs better, but usually both variables do not seem to perform well at the same time. REER series seem to perform almost as well as terms of trade (Figures F7-8).
2. For the CBO U.S. measure of output gap, we confirm the literature following Atkeson and Ohanian (2001), where domestic slack does not help forecast inflation relative to the simple AR process of inflation starting in the mid-1980s (Figures F9-10). In forecasts starting from 1960s through 1970s (where global slack measures are not available), the CBO measure of U.S. slack has a significantly better performance than the AR specification, especially at short horizons. The global slack measures, FRBD G7 and OECD Total, seem to perform only well in forecasting CPI inflation since the late 1980s (during the Great Moderation).
3. The predictive ability of money growth measures (Figures F11-12) is significantly higher when compared to the previous measures considered so far in this paper. With some occasional breaks, the U.S. money supply growth outperforms the naïve forecasts, especially starting in the late 1970s, over long horizons, and across several inflation measures. G7 money supply growth yields even more accurate forecasts over the same sample and horizons than its U.S. counterpart across all inflation measures with the exception of the GDP deflator and PPI inflation where we see a break in the late 1980s. This is interesting because our empirical results based on the benchmark sample are in line with those in D'Agostino and Surico (2009) where they analyze the 1990:Q1-2006:Q2 period and show that global money growth-based forecasts seem to be more accurate than the domestic money which beats the naïve forecasts of inflation. However, the results in the robustness analyses (where we have a larger sample at each window) reveal that these findings do not seem to be robust to sample selection after the 1980s. Therefore, for some measures of inflation, the information content of money is primarily reflected in the global measure of liquidity growth, rather than the domestic measure.
4. Credit growth measures, for both the U.S. and G7 (Figures F13-14), appear to be highly successful forecasting variables for U.S. inflation. Both variables outperform the simple AR process of inflation

accurately (relative to the naïve specification) based on our initial assessments. Hence we restrict our analysis to nominal measures throughout the paper.

and yield low relative MSFEs especially during the mid-1980s. And unlike in the case of the money growth measures, the information content of credit growth is strongly reflected in the global measures rather than in their domestic counterparts. The G7 credit growth yields highly accurate forecasts across all inflation measures, over all horizons after the mid-1980s.

In the next section, we aim to investigate the causes behind these patterns. Some of these variables, such as domestic and global credit, are not present in the model at hand and therefore, we mainly focus on shedding light on the patterns for slack, money and terms of trade.

5 Relating Theory to Stylized Facts

Now we compare and contrast theoretical and empirical results to explain how major episodes for the U.S. economy can be related to the observed forecasting performances reported in this paper. We analyze figures F3-14 paying particular attention to CPI inflation, since the model at hand is primarily aimed at modelling this measure of inflation. However, forecasting patterns are robust to other measures of inflation to a great extent.

The high performance of the Phillips curve-based simulated forecasts is observed at relatively short horizons (i.e. 1- or 4-quarters ahead forecasts). However, what is interesting is that this observation is also valid for the forecasts with money growth measures—both domestic and global. Money growth is incorporated into the model through a money demand equation. In that setting, our model based on the short-run open-economy Phillips curve suggests that money as a proxy for the unobserved output gap should perform well at forecasting. In fact, nominal money growth does not perform well at the 12-quarters ahead forecasts that we consider in the simulations. Based on our empirical results it is interesting to note, however, that money performs well at longer horizons. This observation may be connected with the quantity theory of money (i.e. with a purely monetarist interpretation).²⁹ The model explains the short-run, and at shorter horizons money is a good explanatory variable to forecast inflation because it is a good proxy for the unobserved measures of slack. However, the model does not fully flesh out the long-run implications, so it cannot offer much guidance on that dimension. Still, the fact that it empirically turns out to be the case reinforces the value of these monetary aggregates in forecasting inflation.

These results also show that domestic slack and the terms of trade (first-differenced) serve together as a good proxy for global slack. This would be true for the real exchange rate (first-differenced), which is proportional to the terms of trade in the model, and therefore we do not report those results here. In all experiments, forecasts with domestic slack and terms of trade outperform the naïve forecasts around the benchmark parameterization, at relatively short horizons. This indicates that by first differencing the terms of trade series, we obtain a good proxy for the unobserved, model-consistent terms of trade gap. However, our empirical analysis yields mixed results with a combination of domestic slack (CBO measure) and terms of trade or REER,³⁰ and better results with terms of trade or REER alone. In light of the simulations, we view this as a measurement issue primarily for the output gap observed in the data.

Finally, in order to circumvent the measurement problems to which we alluded before, the current open-economy NK model is particularly useful to show that global slack can be represented by (or proportional

²⁹Also, in a recent study, Sargent and Surico (2011) provide results on the empirical evidence of the quantity theory of money that may explain the performance of U.S. money growth in forecasting U.S. inflation.

³⁰We do not report these results, but they can be provided upon request.

to) not only domestic and foreign slack but also combinations of *i*) domestic slack and the terms of trade gap, and *ii*) domestic and foreign money growth.

6 Conclusion

The seminal work of Atkeson and Ohanian (2001) documented a break in the Phillips curve during the Great Moderation period. This basic statistical relationship between domestic inflation and domestic economic activity no longer seemed to work as a tool for inflation forecasting. Low forecast accuracy can be an issue not only with reduced-form forecasting models, but also with the Dynamic Stochastic General Equilibrium (DSGE) models that have become commonplace for policy analysis and forecasting, as indicated by Edge and Gürkaynak (2010). Focusing on the strand of literature that followed the work of Atkeson and Ohanian (2001), we find theoretical and empirical support for the validity of the global slack hypothesis based on its predictions about forecasting. We show that the Phillips curve is alive and well for forecasting, after all—so long as one considers an open-economy Phillips curve model rather than the standard closed-economy specifications. This is a major contribution of our paper bridging the gap between the theoretical and empirical open-economy literatures.

Based on the open-economy NK model of Martínez-García and Wynne (2010), we specify a broad definition of the open-economy Phillips curve that ties inflation to (i) domestic and foreign slack, (ii) domestic slack and terms of trade/real exchange rate gap, (iii) world money supply growth, where all three specifications involve different ways of embedding global slack into the Phillips curve. In this open-economy NK framework, we establish theoretically that no variable other than global slack helps forecast changes in domestic inflation. We emphasize the importance of alternative measures that incorporate the same information as global slack for forecasting inflation with actual data since it is often a challenging task to find reliable measures of slack purely on measured economic activity. We argue that measures such as global credit growth or money growth, terms of trade and even the real exchange rate have information content about global slack that can be successfully leveraged to forecast domestic inflation—in the case of the U.S., these alternative measures can be more useful than existing measures of U.S. and global slack. Moreover, these variables perform well when compared against forecasts obtained with many conventional, domestic inflation-predictors such as: domestic slack, global slack, domestic money supply growth and domestic credit growth.

Our interpretation of how these variables are linked to inflation, which is derived from an open-economy NK model, has not been considered before. Our quantitative analysis using empirically plausible parameter values of the open-economy NK model show that open-economy Phillips curves have gained more value in forecasting inflation not only through greater openness to trade but also—and to a greater extent—due to changes in the conduct of monetary policy, with a higher anti-inflationary bias over the past few decades, and also due to good luck (i.e. lower exogenous volatility of productivity shocks during the Great Moderation period). We believe that the role of greater openness to trade is defined in a rather limited fashion in our workhorse open-economy NK model and therefore, its impact might be underestimated due to increasing financial globalization and the role of international migration patterns on the labor market. These are avenues to enrich the model that we plan to consider in future research.

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Appendix

A Open-Economy New Keynesian Model

Table A1 - Open-Economy New Keynesian Model: Core Equations

Home Economy	
Phillips curve	$\hat{\pi}_t - \bar{\pi}_t \approx \beta \mathbb{E}_t (\hat{\pi}_{t+1} - \bar{\pi}_{t+1}) + \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) [((1-\xi)\varphi + \Theta\gamma) \hat{x}_t + (\xi\varphi + (1-\Theta)\gamma) \hat{x}_t^*]$
Output gap	$\gamma(1-2\xi)(\mathbb{E}_t [\hat{x}_{t+1}] - \hat{x}_t) \approx ((1-2\xi) + \Gamma) [\hat{r}_t - \hat{r}_t^*] - \Gamma [\hat{r}_t^* - \hat{r}_t^*]$
Monetary policy	$\hat{i}_t \approx \tilde{\pi}_t + [\Psi_\pi (\hat{\pi}_t - \tilde{\pi}_t) + \Psi_x \hat{x}_t] + \hat{v}_t, \tilde{\pi}_t = \tilde{\pi}_{t-1} + \hat{\varepsilon}_t$
Fisher equation	$\hat{r}_t \equiv \hat{i}_t - \mathbb{E}_t [\hat{\pi}_{t+1}]$
Natural interest rate	$\hat{r}_t \approx \gamma [\Theta (\mathbb{E}_t [\hat{y}_{t+1}] - \hat{y}_t) + (1-\Theta) (\mathbb{E}_t [\hat{y}_{t+1}^*] - \hat{y}_t^*)]$
	$\hat{r}_t \approx \gamma \left(\frac{1+\varphi}{\gamma+\varphi} \right) [(\Theta\Lambda + (1-\Theta)(1-\Lambda)) \mathbb{E}_t [\Delta \hat{a}_{t+1}] + (\Theta(1-\Lambda) + (1-\Theta)\Lambda) \mathbb{E}_t [\Delta \hat{a}_{t+1}^*]]$
Potential output	$\hat{y}_t \approx \left(\frac{1+\varphi}{\gamma+\varphi} \right) [\Lambda \hat{a}_t + (1-\Lambda) \hat{a}_t^*]$
Foreign Economy	
Phillips curve	$\hat{\pi}_t^* - \bar{\pi}_t^* \approx \beta \mathbb{E}_t (\hat{\pi}_{t+1}^* - \bar{\pi}_{t+1}^*) + \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) [(\xi\varphi + (1-\Theta)\gamma) \hat{x}_t + ((1-\xi)\varphi + \Theta\gamma) \hat{x}_t^*]$
Output gap	$\gamma(1-2\xi)(\mathbb{E}_t [\hat{x}_{t+1}^*] - \hat{x}_t^*) \approx -\Gamma [\hat{r}_t - \hat{r}_t^*] + ((1-2\xi) + \Gamma) [\hat{r}_t^* - \hat{r}_t^*]$
Monetary policy	$\hat{i}_t^* \approx \tilde{\pi}_t^* + [\psi_\pi (\hat{\pi}_t^* - \tilde{\pi}_t^*) + \psi_x \hat{x}_t^*] + \hat{v}_t^*, \tilde{\pi}_t^* = \tilde{\pi}_{t-1}^* + \hat{\varepsilon}_t^*,$
Fisher equation	$\hat{r}_t^* \equiv \hat{i}_t^* - \mathbb{E}_t [\hat{\pi}_{t+1}^*]$
Natural interest rate	$\hat{r}_t^* \approx \gamma [(1-\Theta) (\mathbb{E}_t [\hat{y}_{t+1}] - \hat{y}_t) + \Theta (\mathbb{E}_t [\hat{y}_{t+1}^*] - \hat{y}_t^*)]$
	$\hat{r}_t^* \approx \gamma \left(\frac{1+\varphi}{\gamma+\varphi} \right) [((1-\Theta)\Lambda + \Theta(1-\Lambda)) \mathbb{E}_t [\Delta \hat{a}_{t+1}] + ((1-\Theta)(1-\Lambda) + \Theta\Lambda) \mathbb{E}_t [\Delta \hat{a}_{t+1}^*]]$
Potential output	$\hat{y}_t^* \approx \left(\frac{1+\varphi}{\gamma+\varphi} \right) [(1-\Lambda) \hat{a}_t + \Lambda \hat{a}_t^*]$
Exogenous, Country-Specific Shocks	
Productivity shock	$\begin{pmatrix} \hat{a}_t \\ \hat{a}_t^* \end{pmatrix} \approx \begin{pmatrix} \delta_a & \delta_{a,a^*} \\ \delta_{a,a^*} & \delta_a \end{pmatrix} \begin{pmatrix} \hat{a}_{t-1} \\ \hat{a}_{t-1}^* \end{pmatrix} + \begin{pmatrix} \hat{\varepsilon}_t^a \\ \hat{\varepsilon}_t^{a^*} \end{pmatrix}$
	$\begin{pmatrix} \hat{\varepsilon}_t^a \\ \hat{\varepsilon}_t^{a^*} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_a^2 & \rho_{a,a^*} \sigma_a^2 \\ \rho_{a,a^*} \sigma_a^2 & \sigma_a^2 \end{pmatrix} \right)$
Monetary shock	$\begin{pmatrix} \hat{v}_t \\ \hat{v}_t^* \end{pmatrix} \approx \begin{pmatrix} \delta_v & 0 \\ 0 & \delta_v \end{pmatrix} \begin{pmatrix} \hat{v}_{t-1} \\ \hat{v}_{t-1}^* \end{pmatrix} + \begin{pmatrix} \hat{\varepsilon}_t^v \\ \hat{\varepsilon}_t^{v^*} \end{pmatrix}$
	$\begin{pmatrix} \hat{\varepsilon}_t^v \\ \hat{\varepsilon}_t^{v^*} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_v^2 & \rho_{v,v^*} \sigma_v^2 \\ \rho_{v,v^*} \sigma_v^2 & \sigma_v^2 \end{pmatrix} \right)$
Composite Parameters	
	$\Theta \equiv (1-\xi) \left[\frac{\sigma\gamma - (\sigma\gamma-1)(1-2\xi)}{\sigma\gamma - (\sigma\gamma-1)(1-2\xi)^2} \right]$
	$\Lambda \equiv 1 + (\sigma\gamma-1) \left[\frac{\gamma\xi 2(1-\xi)}{\varphi(\sigma\gamma - (\sigma\gamma-1)(1-2\xi)^2) + \gamma} \right]$
	$\Gamma \equiv \xi [\sigma\gamma + (\sigma\gamma-1)(1-2\xi)]$

Table A2 - Open-Economy New Keynesian Model: Non-Core Equations	
Home Economy	
Output	$\hat{y}_t = \hat{\bar{y}}_t + \hat{x}_t$
Consumption	$\hat{c}_t \approx \Theta \hat{y}_t + (1 - \Theta) \hat{y}_t^*$
Employment	$\hat{l}_t \approx \hat{y}_t - \hat{a}_t$
Real wages	$(\hat{w}_t - \hat{p}_t) \approx \gamma \hat{c}_t + \phi \hat{l}_t \approx (\varphi + \gamma \Theta) \hat{y}_t + \gamma (1 - \Theta) \hat{y}_t^* - \varphi \hat{a}_t$
Real Money Demand	$\hat{m}_t^d - \hat{p}_t \approx \hat{c}_t - \eta \hat{i}_t$
Foreign Economy	
Output	$\hat{y}_t^* = \hat{\bar{y}}_t^* + \hat{x}_t^*$
Consumption	$\hat{c}_t^* \approx (1 - \Theta) \hat{y}_t + \Theta \hat{y}_t^*$
Employment	$\hat{l}_t^* \approx \hat{y}_t^* - \hat{a}_t^*$
Real wages	$(\hat{w}_t^* - \hat{p}_t^*) \approx \gamma \hat{c}_t^* + \phi \hat{l}_t^* \approx \gamma (1 - \Theta) \hat{y}_t + (\varphi + \gamma \Theta) \hat{y}_t^* - \varphi \hat{a}_t^*$
Real Money Demand	$\hat{m}_t^{d*} - \hat{p}_t^* \approx \hat{c}_t^* - \eta \hat{i}_t^*$
International Relative Prices and Trade	
Real exchange rate	$\hat{r}_{\$t} \approx (1 - 2\xi) \text{tot}_t$
Terms of trade	$\widehat{\text{tot}}_t \approx \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\xi)^2} \right] (\hat{y}_t - \hat{y}_t^*)$
Home real exports	$\widehat{\text{exp}}_t \approx \Xi \hat{y}_t + (1 - \Xi) \hat{y}_t^*$
Home real imports	$\widehat{\text{imp}}_t \approx - (1 - \Xi) \hat{y}_t - \Xi \hat{y}_t^*$
Home real trade balance	$\hat{tb}_t \equiv \hat{y}_t - \hat{c}_t = (1 - \xi) \left(\widehat{\text{exp}}_t - \widehat{\text{imp}}_t \right) \approx (1 - \Theta) (\hat{y}_t - \hat{y}_t^*)$
Composite Parameters	
$\Theta \equiv (1 - \xi) \left[\frac{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\xi)}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\xi)^2} \right]$	
$\Xi \equiv \left[\frac{\sigma\gamma + (\sigma\gamma - 1)(1 - 2\xi)\xi}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\xi)^2} \right]$	

B Inflation Dynamics in the Open-Economy New Keynesian Model

In order to clarify the country-level dynamics of the open-economy model in Martínez-García and Wynne (2010) that we described in section 3, we give the details on the dynamics of the global and difference economy. In particular, we use the decomposition method into aggregates and differences advocated by Aoki (1981) and Fukuda (1993) to re-express the core linear rational expectations system that characterizes the log-linearized solution into two separate sub-systems. Productivity shocks enter into the dynamics of the model only through their impact on the dynamics of the natural (real) rates in this economy, \hat{r}_t and \hat{r}_t^* . The Home and Foreign monetary shock processes \hat{v}_t and \hat{v}_t^* enter through the specification of the Taylor monetary policy rule of each country and capture the central bank's expected real rate.

The two countries are assumed to be symmetric in every respect, except on their consumption basket due to the assumption of home-product bias in consumption. Even so, the specification of the home-product bias is inherently symmetric as well since the share of local goods in the local consumption basket is the same in both countries and determined by the parameter ζ . Hence, we define the world aggregate and the difference variables \hat{g}_t^W and \hat{g}_t^R as,

$$\hat{g}_t^W \equiv \frac{1}{2}\hat{g}_t + \frac{1}{2}\hat{g}_t^*, \quad (30)$$

$$\hat{g}_t^R \equiv \hat{g}_t - \hat{g}_t^*, \quad (31)$$

which implicitly takes into account that both countries are identical in size (with the same share of the household population and varieties located in each country). We re-write the country variables \hat{g}_t and \hat{g}_t^* as,

$$\hat{g}_t = \hat{g}_t^W + \frac{1}{2}\hat{g}_t^R, \quad (32)$$

$$\hat{g}_t^* = \hat{g}_t^W - \frac{1}{2}\hat{g}_t^R. \quad (33)$$

If we characterize the dynamics for \hat{g}_t^W and \hat{g}_t^R , the transformation above backs out the corresponding variables for each country \hat{g}_t and \hat{g}_t^* . Naturally, these transformations can be applied to any of the endogenous and exogenous variables in the model. Then, under this transformation, we can orthogonalize the original two-country model of Martínez-García and Wynne (2010) into one aggregate (or world) economic system and one difference system that can be studied independently.

The World Economy. The global system describes the world economy, as if it were a closed economy, based on the following system of three equations (in addition to the equality of inflation trend and target for all periods, i.e. $\bar{\pi}_t^W = \tilde{\pi}_t^W$)

$$\hat{\pi}_t^W - \bar{\pi}_t^W = \beta \mathbb{E}_t \left[\hat{\pi}_{t+1}^W - \bar{\pi}_{t+1}^W \right] + \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) (\varphi + \gamma) \hat{x}_t^W, \quad (34)$$

$$\gamma \left(\mathbb{E}_t \left[\hat{x}_{t+1}^W \right] - \hat{x}_t^W \right) \approx \left(\hat{i}_t^W - \mathbb{E}_t \left[\hat{\pi}_{t+1}^W \right] \right) - \hat{r}_t^W, \quad (35)$$

$$\hat{i}_t^W = \bar{\pi}_t^W + \psi_\pi \left(\hat{\pi}_t^W - \bar{\pi}_t^W \right) + \hat{v}_t^W, \quad (36)$$

which can be expressed more compactly in two equations as,

$$\hat{\pi}_t^W - \bar{\pi}_t^W \approx \beta \mathbb{E}_t \left[\hat{\pi}_{t+1}^W - \bar{\pi}_{t+1}^W \right] + \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) (\varphi + \gamma) \hat{x}_t^W, \quad (37)$$

$$\gamma \mathbb{E}_t \left[\hat{x}_{t+1}^W \right] - (\gamma + \psi_x) \hat{x}_t^W \approx \bar{\pi}_t^W + \left(\psi_\pi \left(\hat{\pi}_t^W - \bar{\pi}_t^W \right) - \mathbb{E}_t \left[\hat{\pi}_{t+1}^W \right] \right) + \left(\hat{v}_t^W - \hat{r}_t^W \right). \quad (38)$$

To close the world economy system, we derive the world forcing processes \hat{r}_t^W and \hat{v}_t^W as follows,

Lemma 1 *Given the derivation of the natural rates for each country and the assumptions on the monetary shock, the world forcing processes for \hat{r}_t^W and \hat{v}_t^W can be described as follows,*

$$\begin{pmatrix} \hat{r}_t^W \\ \hat{v}_t^W \end{pmatrix} = \begin{pmatrix} \delta_a + \delta_{a,a^*} & 0 \\ 0 & \delta_m \end{pmatrix} \begin{pmatrix} \hat{r}_{t-1}^W \\ \hat{v}_{t-1}^W \end{pmatrix} + \begin{pmatrix} \hat{\varepsilon}_t^{rW} \\ \hat{\varepsilon}_t^{vW} \end{pmatrix}, \quad (39)$$

$$\begin{pmatrix} \hat{\varepsilon}_t^{rW} \\ \hat{\varepsilon}_t^{vW} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_r^2 \left(\frac{1+\rho_{r,r^*}}{2} \right) & 0 \\ 0 & \sigma_v^2 \left(\frac{1+\rho_{v,v^*}}{2} \right) \end{pmatrix} \right), \quad (40)$$

where the volatility term for the world natural rate can be tied to parameters of the productivity shock and other structural parameters of the model as,

$$\sigma_r^2 \left(\frac{1+\rho_{r,r^*}}{2} \right) = \sigma_a^2 \left(\frac{1+\rho_{a,a^*}}{2} \right) \left[\gamma \left(\frac{1+\varphi}{\gamma+\varphi} \right) (\delta_a + \delta_{a,a^*} - 1) \right]^2. \quad (41)$$

The Difference Economy. The difference system defines how far apart each country is from the other. Then, the difference system can be characterized as follows,

$$\hat{\pi}_t^R - \bar{\pi}_t^R \approx \beta \mathbb{E}_t \left[\hat{\pi}_{t+1}^R - \bar{\pi}_{t+1}^R \right] + \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) ((1-2\zeta)\varphi + (2\Theta-1)\gamma) \hat{x}_t^R, \quad (42)$$

$$\gamma(1-2\zeta) \left(\mathbb{E}_t \left[\hat{x}_{t+1}^R \right] - \hat{x}_t^R \right) \approx ((1-2\zeta) + 2\Gamma) \left[\left(\hat{\pi}_t^R - \mathbb{E}_t \left[\hat{\pi}_{t+1}^R \right] \right) - \hat{r}_t^R \right], \quad (43)$$

$$\hat{i}_t^R \approx \bar{\pi}_t^R + \psi_\pi \left(\hat{\pi}_t^R - \bar{\pi}_t^R \right) + \psi_x \hat{x}_t^R + \hat{v}_t^R, \quad (44)$$

which can be expressed more compactly in two equations as,

$$\hat{\pi}_t^R - \bar{\pi}_t^R \approx \beta \mathbb{E}_t \left(\hat{\pi}_{t+1}^R - \bar{\pi}_{t+1}^R \right) + \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) ((1-2\zeta)\varphi + (2\Theta-1)\gamma) \hat{x}_t^R, \quad (45)$$

$$\begin{aligned} & \gamma(1-2\zeta) \mathbb{E}_t \left[\hat{x}_{t+1}^R \right] - ((\gamma + \psi_x)(1-2\zeta) + 2\psi_x\Gamma) \hat{x}_t^R \\ & \approx ((1-2\zeta) + 2\Gamma) \left[\bar{\pi}_t^R + \psi_\pi \left(\hat{\pi}_t^R - \bar{\pi}_t^R \right) - \mathbb{E}_t \left[\hat{\pi}_{t+1}^R \right] + \left(\hat{v}_t^R - \hat{r}_t^R \right) \right]. \end{aligned} \quad (46)$$

Here, the degree of openness ζ plays an important role in the difference system of equations and so does the elasticity of substitution between Home and Foreign consumption bundles σ (through the composite parameters Θ and Γ).

To close the difference economy system, we derive the difference forcing processes \hat{r}_t^R and \hat{v}_t^R as follows,

Lemma 2 *Given the derivation of the natural rates for each country and the assumptions on the monetary shocks,*

the difference forcing processes for \hat{r}_t^R and \hat{v}_t^R can be described as follows,

$$\begin{pmatrix} \hat{r}_t^R \\ \hat{v}_t^R \end{pmatrix} = \begin{pmatrix} \delta_a - \delta_{a,a^*} & 0 \\ 0 & \delta_v \end{pmatrix} \begin{pmatrix} \hat{r}_{t-1}^R \\ \hat{v}_{t-1}^R \end{pmatrix} + \begin{pmatrix} \hat{\varepsilon}_t^{rR} \\ \hat{\varepsilon}_t^{vR} \end{pmatrix}, \quad (47)$$

$$\begin{pmatrix} \hat{\varepsilon}_t^{rR} \\ \hat{\varepsilon}_t^{vR} \end{pmatrix} \sim N \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 2\sigma_r^2(1 - \rho_{r,r^*}) & 0 \\ 0 & 2\sigma_v^2(1 - \rho_{v,v^*}) \end{pmatrix} \right), \quad (48)$$

where the volatility term for the difference natural rate can be tied to parameters of the productivity shock and other structural parameters of the model as,

$$2\sigma_r^2(1 - \rho_{r,r^*}) = 2\sigma_a^2(1 - \rho_{a,a^*}) \left[\gamma \left(\frac{1 + \varphi}{\gamma + \varphi} \right) (2\Theta - 1)(2\Lambda - 1)(\delta_a - \delta_{a,a^*} - 1) \right]^2. \quad (49)$$

B.1 Dynamics of World Inflation

We build our empirical model on the basis of the work on global slack of Martínez-García and Wynne (2010) and Martínez-García (2014). In their model the world economy can be described with three equations as can be seen in (34) – (36) that have the same basic structure as one would find in the standard three-equation, closed-economy NK model.

The world economy NK model of Martínez-García (2014) is described with a New Keynesian Phillips curve (NKPC), a log-linearized world Euler equation, and an interest-rate-setting rule for monetary policy. The NKPC can be cast into the following augmented form,

$$\hat{\pi}_t^W - \bar{\pi}_t^W = \beta \mathbb{E}_t \left(\hat{\pi}_{t+1}^W - \bar{\pi}_{t+1}^W \right) + k^W \hat{x}_t^W, \quad (50)$$

where $\mathbb{E}_t(\cdot)$ refers to the expectation formed conditional on information up to time t , \hat{x}_t^W is the global output gap, $\hat{\pi}_t^W$ is global inflation, and $\bar{\pi}_t^W$ is the global trend inflation. Moreover, $k^W \equiv \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) (\varphi + \gamma) > 0$ is the slope of the global output gap that depends on the deep structural parameters of the model such as the frequency of price adjustment $0 < \alpha < 1$, and the intertemporal discount rate $0 < \beta < 1$. The NKPC describing the dynamics of aggregate world inflation arises in a two-country model with staggered price-setting à la Calvo (1983), and can be augmented to include a time-varying global trend for inflation with a standard extension to incorporate price indexation in the price-setting decision of firms as in Yun (1996). In such an environment, firms that do not re-optimize their prices would automatically increase them at the trend inflation rate of the county where they reside.

The log-linearization of the Euler equation is given by,

$$\hat{x}_t^W = \mathbb{E}_t \left[\hat{x}_{t+1}^W \right] - \frac{1}{\gamma} \left(\hat{i}_t^W - \mathbb{E}_t \left[\hat{\pi}_{t+1}^W \right] - \hat{r}_t^W \right), \quad (51)$$

where \hat{i}_t^W is the aggregate short-term nominal interest rate (an aggregate of the riskless one-period interest rates of both countries), and \hat{r}_t^W is the aggregate natural interest rate—the real interest rate that the economy would have experienced absent nominal rigidities, given the same realization of the real shocks. Potential output and the natural (real) interest rate are both functions of exogenous real factors (technology).

We specify a general form of the monetary policy with a Taylor (1993) rule where the central bank of each country targets their domestic short-term nominal interest rate with the same reaction function. The world Taylor rule can be cast in the following form,

$$\hat{i}_t^W = \tilde{\pi}_t^W + \psi_\pi (\hat{\pi}_t^W - \tilde{\pi}_t^W) + \psi_x \hat{x}_t^W + \hat{v}_t^W, \quad (52)$$

where $\tilde{\pi}_t^W$ is the aggregate of both countries' central bank's inflation target and \hat{r}_t^W can be interpreted as the aggregate of each country's central bank's own forecast of the economy's natural (real) interest rate. We assume that the inflation target for each country follows a random walk so that the aggregate itself, $\tilde{\pi}_t^W$, also follows a random walk, i.e.

$$\tilde{\pi}_t^W = \tilde{\pi}_{t-1}^W + \tilde{\varepsilon}_t, \quad (53)$$

where $\tilde{\varepsilon}_t$ is an i.i.d. shock with zero mean. In our implementation with simulated data we collapse the target rate to a constant and normalize it to zero to be consistent with the simple scenario of a zero inflation steady state. However, we maintain the generality of the specification here for illustration purposes.

In this setting, the aggregate trend inflation $\bar{\pi}_t^W$ corresponds in equilibrium to the aggregate of the central bank's inflation target $\tilde{\pi}_t^W$. To see that, one can interpret the aggregate indexation rate $\bar{\pi}_t^W$ as the Beveridge-Nelson (stochastic) trend of the global inflation process,

$$\bar{\pi}_t^W = \lim_{j \rightarrow \infty} \mathbb{E}_t (\hat{\pi}_{t+j}^W). \quad (54)$$

The world inflation rate $\hat{\pi}_t^W$ in this model fluctuates around a stochastic trend given by the aggregate central bank's inflation target. Hence, since we assume in (53) that the target is a random walk, it follows that $\mathbb{E}_t (\hat{\pi}_{t+j}^W) = \tilde{\pi}_t^W$ at any period $j > 0$. In that case, it results from the definition in (54) that $\bar{\pi}_t^W = \tilde{\pi}_t^W$ at every point in time and this confirms that trend and target inflation must be equal in equilibrium.

Using the aggregate monetary policy rule in (52) to replace \hat{i}_t^W in (50) – (51), the system of equations that determines world inflation and global slack can be written in the following form,

$$\hat{z}_t^W = A^W \mathbb{E}_t (\hat{z}_{t+1}^W) + a^W (\hat{r}_t^W - \hat{v}_t^W), \quad (55)$$

where,

$$\hat{z}_t^W \equiv \begin{bmatrix} \hat{\pi}_t^W - \bar{\pi}_t^W \\ \hat{x}_t^W \end{bmatrix}, \quad (56)$$

where A^W is a 2×2 matrix and a^W is a 2×1 matrix of structural coefficients. We assume that the process for the aggregate central bank's predicted real rate \hat{r}_t^W is stochastic and exogenous. Under the assumption that the aggregate interest rate gap $(\hat{r}_t^W - \hat{v}_t^W)$ is stationary, then the system in (55) has a unique nonexplosive solution in which both \hat{x}_t^W and $\hat{\pi}_t^W - \bar{\pi}_t^W$ are stationary whenever both eigenvalues of the matrix A^W are inside the unit circle. A variant of the Taylor principle which requires that $\psi_\pi + \left(\frac{1-\beta}{k^W}\right) \psi_x > 1$ suffices to ensure the uniqueness and existence of the nonexplosive solution for the world aggregates. Assuming this

condition is satisfied, the solution can be characterized as follows,

$$\begin{pmatrix} \hat{\pi}_t^W \\ \hat{x}_t^W \end{pmatrix} = \begin{pmatrix} \bar{\pi}_t^W \\ 0 \end{pmatrix} + \sum_{j=0}^{\infty} (A^W)^j a^j \mathbb{E}_t (\hat{r}_{t+j}^W - \hat{v}_{t+j}^W). \quad (57)$$

Hence, world inflation is determined by the world inflation target and by current and expected future discrepancies between the aggregate natural rate of interest and the aggregate of the central bank's own target for the natural rate.

We assume that the central banks adjust their policy rule to track changes in the natural rate of interest that are forecastable one period in advance implying for the aggregate that,

$$\hat{v}_t^W = \mathbb{E}_{t-1} (\hat{r}_t^W). \quad (58)$$

Alternatively, we can simply assume—as most of the literature implicitly does—that $\hat{v}_t^W = \hat{r}_t^W + \hat{\varepsilon}_t^m$, where \hat{r}_t^W corresponds to the global natural interest rate and $\hat{\varepsilon}_t^m$ is an i.i.d. disturbance that captures non-persistent and unanticipated shocks to monetary policy. In either case, the world interest rate gap $(\hat{r}_t^W - \hat{v}_t^W)$ is viewed as white noise and the solution to the global system in (55) becomes,

$$\hat{\pi}_t^W = \bar{\pi}_t^W + \lambda^W (\hat{r}_t^W - \hat{v}_t^W) = \bar{\pi}_t^W - \lambda^W \hat{\varepsilon}_t^m, \quad (59)$$

$$\hat{x}_t^W = \mu^W (\hat{r}_t^W - \hat{v}_t^W) = -\mu^W \hat{\varepsilon}_t^m, \quad (60)$$

where the composite coefficients λ^W and μ^W depend on the deep structural parameters of the model. If aggregate inflation evolves as predicted by this solution, then optimal forecasts of future global inflation at any horizon $j \geq 1$ must be given by,

$$\mathbb{E}_t (\hat{\pi}_{t+j}^W) = \bar{\pi}_t^W = \hat{\pi}_t^W - \frac{\lambda^W}{\mu^W} \hat{x}_t^W, \quad (61)$$

or, simply re-arranging, by,

$$\mathbb{E}_t (\hat{\pi}_{t+j}^W - \hat{\pi}_t^W) = -\frac{\lambda^W}{\mu^W} \hat{x}_t^W. \quad (62)$$

More generally, using the fact that $\hat{\pi}_{t+h|t}^{W,h} \approx \frac{400}{h} \sum_{j=1}^h \hat{\pi}_{t+j}^W$, we can write the forecast h -periods ahead as follows,

$$\mathbb{E}_t (\hat{\pi}_{t+h|t}^{W,h}) = \frac{400}{h} \sum_{j=1}^h \mathbb{E}_t (\hat{\pi}_{t+j}^W) = \frac{400}{h} \sum_{j=1}^h \mathbb{E}_t (\bar{\pi}_{t+j}^W) = 400 \left(\hat{\pi}_t^W - \frac{\lambda^W}{\mu^W} \hat{x}_t^W \right). \quad (63)$$

This implies that no other variable should improve our forecast of changes in the global inflation if global slack and the current global inflation rate are included in the forecasting model. This feature is noted in Woodford (2008) as well, and we use it as our key identifying restriction in order to construct a reduced-form specification (an ADL model) for forecasting inflation that is consistent with the NKPC.

Forecasting future global inflation using the global output gap alone would not be accurate since global inflation potentially has a stochastic trend while global slack is stationary; one needs to include among the

regressors some variable with a similar stochastic trend to that of inflation. But this need not be money growth; current global inflation itself has the same stochastic trend, so including it to forecast future inflation takes care of the trend component without the need to include any other regressors to attempt to track the stochastic trend.

What we need apart from current global inflation is additional regressors that are stationary and highly correlated with the current deviations of inflation from its stochastic trend. In theory, the global output gap is one such stationary variable with that property. More generally, what matters is which variables are most useful for tracking relatively high-frequency (or cyclical) variations in inflation. This is true regardless of the horizon over which one wishes to forecast inflation. In this sense, we find that global money can be a relevant variable to help us forecast inflation.

Proposition 2 *World real money gap $\hat{m}_t^{s,W}$ is proportional to global slack,*

$$\hat{m}_t^{s,W} \approx \chi \hat{x}_t^W, \quad (64)$$

where $\chi \equiv \left(1 - \eta \left(-\psi_\pi \frac{\lambda^W}{\mu^W} + \psi_x\right)\right)$.

Proof. The aggregate money demand equations can be expressed as follows,

$$\hat{m}_t^{d,W} - \hat{p}_t^W \approx \hat{c}_t^W - \eta \hat{i}_t^W, \quad (65)$$

where aggregate world consumption is given by $\hat{c}_t^W \approx \hat{y}_t^W$. Under the solution described here and the implication that the global inflation trend and the aggregate inflation target for the central banks must equate, we know that the aggregate Taylor rule implies the following path for the nominal short-term interest rate,

$$\hat{i}_t^W = \bar{\pi}_t^W + \left(-\psi_\pi \frac{\lambda^W}{\mu^W} + \psi_x\right) \hat{x}_t^W + \hat{v}_t^W. \quad (66)$$

When we express the counterpart of the aggregate money demand in (65) absent nominal rigidities, we use the fact that $\hat{i}_t^W = \bar{\pi}_t^W + \hat{v}_t^W$ and write it as follows,

$$\begin{aligned} \hat{m}_t^{s,W} &\equiv \left(\hat{m}_t^{d,W} - \hat{p}_t^W\right) - \left(\hat{m}_t^{d,W} - \hat{p}_t^W\right) \approx \left(\hat{c}_t^W - \hat{c}_t^W\right) - \eta \left(\hat{i}_t^W - \hat{i}_t^W\right) \\ &\approx \left(\hat{y}_t^W - \hat{y}_t^W\right) - \eta \left(\hat{i}_t^W - \hat{i}_t^W\right) \\ &\approx \left(1 - \eta \left(-\psi_\pi \frac{\lambda^W}{\mu^W} + \psi_x\right)\right) \hat{x}_t^W. \end{aligned} \quad (67)$$

■

In other words, the gap in real money demand ought to be proportional to global slack and, therefore, can be used for forecasting. Similarly, we will look at the terms of trade to obtain information that could be helpful for forecasting inflation in place of the global slack.

Under a Taylor rule specification such as the one we adopt here, the price level is known to be indeterminate even when a solution for the model on the inflation rate can be shown to exist and be unique (see, e.g., Cochrane (2011)). Using the aggregate money demand equation and the aggregate Fisher equation,

we can write the following expression in terms of the price level,

$$\begin{aligned}\hat{m}_t^{d,W} - \hat{p}_t^W &\approx \hat{y}_t^W - \eta \left(\hat{r}_t^W + \mathbb{E}_t \left[\hat{\pi}_{t+1}^W \right] \right) \\ &= \hat{y}_t^W - \eta \left(\hat{r}_t^W + \mathbb{E}_t \left[\hat{p}_{t+1}^W \right] - \hat{p}_t^W \right),\end{aligned}\quad (68)$$

or, simply,

$$\hat{p}_t^W \approx \frac{\eta}{1+\eta} \mathbb{E}_t \left[\hat{p}_{t+1}^W \right] - \frac{1}{1+\eta} \left(\hat{y}_t^W - \eta \hat{r}_t^W \right) + \frac{1}{1+\eta} \hat{m}_t^{d,W}. \quad (69)$$

Solving it forwards, we obtain the following expression under the no-bubbles assumption (i.e. $\lim_{T \rightarrow \infty} \left(\frac{\eta}{1+\eta} \right)^T \mathbb{E}_t \left[\hat{p}_{t+T}^W \right] = 0$),

$$\hat{p}_t^W \approx \sum_{k=0}^{\infty} \left(\frac{\eta}{1+\eta} \right)^k \mathbb{E}_t \left[-\frac{1}{1+\eta} \left(\hat{y}_{t+k}^W - \eta \hat{r}_{t+k}^W \right) \right] + \sum_{k=0}^{\infty} \left(\frac{\eta}{1+\eta} \right)^k \mathbb{E}_t \left[\frac{1}{1+\eta} \hat{m}_{t+k}^{d,W} \right]. \quad (70)$$

Given the derivation of the natural rates and potential output for each country, the world processes for the corresponding aggregates \hat{r}_t^W and \hat{y}_t^W can be described as follows,

$$\hat{r}_t^W = (\delta_a + \delta_{a,a^*}) \hat{r}_{t-1}^W + \hat{\varepsilon}_t^{r,W}, \quad \hat{\varepsilon}_t^{r,W} \sim N \left(0, \sigma_a^2 \left(\frac{1 + \rho_{a,a^*}}{2} \right) \left[\gamma \left(\frac{1 + \varphi}{\gamma + \varphi} \right) (\delta_a + \delta_{a,a^*} - 1) \right]^2 \right), \quad (71)$$

$$\hat{y}_t^W = (\delta_a + \delta_{a,a^*}) \hat{y}_{t-1}^W + \hat{\varepsilon}_t^{y,W}, \quad \hat{\varepsilon}_t^{y,W} \sim N \left(0, \sigma_a^2 \left(\frac{1 + \rho_{a,a^*}}{2} \right) \left[\frac{1 + \varphi}{\gamma + \varphi} \right]^2 \right). \quad (72)$$

Hence, the expression for the global price level can be re-written as follows,

$$\hat{p}_t^W \approx \frac{1}{1+\eta} \left(\hat{y}_t^W - \eta \hat{r}_t^W \right) - \frac{1}{1+\eta} \left(\frac{\eta (\delta_a + \delta_{a,a^*})}{1 + \eta (1 - (\delta_a + \delta_{a,a^*}))} \right) \left(\hat{y}_t^W - \eta \hat{r}_t^W \right) + \sum_{k=0}^{\infty} \left(\frac{\eta}{1+\eta} \right)^k \mathbb{E}_t \left[\frac{1}{1+\eta} \hat{m}_{t+k}^{d,W} \right], \quad (73)$$

where,

$$\begin{aligned}\hat{y}_t^W &= \hat{y}_t^W + \mu^W \left(\hat{r}_t^W - \hat{m}_t^W \right) = \hat{y}_t^W + \hat{x}_t^W, \\ \hat{r}_t^W &\equiv \hat{r}_t^W - \mathbb{E}_t \left[\hat{\pi}_{t+1}^W \right] = \left(-\psi_\pi \frac{\lambda^W}{\mu^W} + \psi_x \right) \hat{x}_t^W + \hat{m}_t^W \\ &= \hat{r}_t^W - \left(1 - \left(-\psi_\pi \frac{\lambda^W}{\mu^W} + \psi_x \right) \mu^W \right) \left(\hat{r}_t^W - \hat{m}_t^W \right) \\ &= \hat{r}_t^W - \left(1 - \left(-\psi_\pi \frac{\lambda^W}{\mu^W} + \psi_x \right) \mu^W \right) \frac{1}{\mu^W} \hat{x}_t^W.\end{aligned}$$

Simple manipulations allow us to write the difference equation for the price level more compactly as follows,

$$\hat{p}_t^W \approx \vartheta_1 \hat{x}_t^W + \vartheta_2 \left(\hat{y}_t^W - \eta \hat{r}_t^W \right) + \sum_{k=0}^{\infty} \left(\frac{\eta}{1+\eta} \right)^k \mathbb{E}_t \left[\frac{1}{1+\eta} \hat{m}_{t+k}^{d,W} \right], \quad (74)$$

where ϑ_1 and ϑ_2 are composite coefficients derived from the structural parameters of the model. Notice that the expression $\vartheta_1 \hat{x}_t^W + \vartheta_2 \left(\hat{y}_t^W - \eta \hat{r}_t^W \right)$ is a weighted sum of two processes that are stationary (one is a

white noise while the other is an autoregressive process of order one).

Adopting the representation proposed by Campbell and Shiller (1987) for this present-value formula, the price level can be expressed alternatively as,

$$\hat{p}_t^W \approx \vartheta_1 \hat{x}_t^W + \vartheta_2 \left(\hat{y}_t^W - \eta \hat{r}_t^W \right) + \sum_{k=1}^{\infty} \left(\frac{\eta}{1+\eta} \right)^k \mathbb{E}_t \left[\Delta \hat{m}_{t+k}^{d,W} \right] + \hat{m}_t^{d,W}, \quad (75)$$

which can be re-expressed in terms of the real demand for money balances as,

$$\hat{m}_t^{d,W} - \hat{p}_t^W \approx -\vartheta_1 \hat{x}_t^W - \vartheta_2 \left(\hat{y}_t^W - \eta \hat{r}_t^W \right) - \sum_{k=1}^{\infty} \left(\frac{\eta}{1+\eta} \right)^k \mathbb{E}_t \left[\Delta \hat{m}_{t+k}^{d,W} \right]. \quad (76)$$

Although the price level is indeterminate, this equation constraints the path of the money demand for any given price level. The way in which it has been re-written indicates that the money demand in real terms must reflect all discounted expected future additions to the stock of money apart from current factors affecting such as \hat{x}_t^W , \hat{y}_t^W and \hat{r}_t^W .

B.2 Dynamics of Differential Inflation

We build our empirical model on the basis of the work on global slack of Martínez-García and Wynne (2010) and Martínez-García (2014). In their model the difference economy can be described with three equations as can be seen in (42) – (44) that have the same basic structure as one would find in the standard three-equation, closed-economy New Keynesian (NK) model.

The difference economy NK model of Martínez-García (2014) is described with a New Keynesian Phillips curve (NKPC), a log-linearized world Euler equation, and an interest-rate-setting rule for monetary policy. The NKPC can be cast into the following augmented form,

$$\hat{\pi}_t^R - \bar{\pi}_t^R = \beta \mathbb{E}_t \left(\hat{\pi}_{t+1}^R - \bar{\pi}_{t+1}^R \right) + k^R \hat{x}_t^R, \quad (77)$$

where $\mathbb{E}_t(\cdot)$ refers to the expectation formed conditional on information up to time t , \hat{x}_t^R is the difference in the current output gap between the two countries, $\hat{\pi}_t^R$ is the difference in inflation, and $\bar{\pi}_t^R$ is the difference in trend inflation. Moreover, $k^R \equiv \left(\frac{(1-\alpha)(1-\beta\alpha)}{\alpha} \right) ((1-2\zeta)\varphi + (2\Theta-1)\gamma)$ is the slope of the difference output gap that depends on the deep structural parameters of the model such as the frequency of price adjustment $0 < \alpha < 1$, and the intertemporal discount rate $0 < \beta < 1$. The NKPC describing the dynamics of the difference in inflation arises in a two-country model with staggered price-setting à la Calvo (1983) and can be augmented to include a time-varying global trend for inflation with a standard extension to incorporate price indexation in the price-setting decision of firms as in Yun (1996). In such an environment, firms that do not re-optimize their prices would automatically increase them at the trend inflation rate of the country where they reside.

The log-linearization of the Euler equation is given by,

$$\hat{x}_t^R = \mathbb{E}_t \left[\hat{x}_{t+1}^R \right] - \frac{1}{\gamma} \left(\frac{(1-2\zeta) + 2\Gamma}{1-2\zeta} \right) \left(\hat{i}_t^R - \mathbb{E}_t \left[\hat{\pi}_{t+1}^R \right] - \hat{r}_t^R \right), \quad (78)$$

where \hat{i}_t^R is the difference in the short-term nominal interest rate (the difference between the riskless one-period interest rates of each country), and \hat{r}_t^R is the difference natural interest rate—the real interest rate differential that the economy would have experienced absent nominal rigidities, given the same realization of the real shocks. Potential output and the natural (real) interest rate are both functions of exogenous real factors (technology).

We specify a general form of the monetary policy with a Taylor (1993) rule where the central bank of each country targets their domestic short-term nominal interest rate with the same reaction function. The world Taylor rule can be cast in the following form,

$$\hat{i}_t^R = \tilde{\pi}_t^R + \psi_\pi (\hat{\pi}_t^R - \tilde{\pi}_t^R) + \psi_x \hat{x}_t^R + \hat{v}_t^R, \quad (79)$$

where $\tilde{\pi}_t^R$ is the difference between both countries' central bank's inflation target and \hat{r}_t^R can be interpreted as the difference between both country's central bank's own forecast of the economy's natural (real) interest rate. We assume that the inflation target for each country follows a random walk so that the difference itself, $\tilde{\pi}_t^R$, also follows a random walk, i.e.

$$\tilde{\pi}_t^R = \tilde{\pi}_{t-1}^R + \tilde{\varepsilon}_t^*, \quad (80)$$

where $\tilde{\varepsilon}_t^*$ is an i.i.d. shock with zero mean. In our implementation with simulated data we collapse the target rate to a constant and normalize it to zero to be consistent with the simple scenario of a zero inflation steady state. However, we maintain the generality of the specification here for illustration purposes.

In this setting, it also follows that the difference trend inflation $\bar{\pi}_t^R$ corresponds in equilibrium to the difference of the central bank's inflation target $\tilde{\pi}_t^R$. To see that, one can interpret the aggregate indexation rate $\bar{\pi}_t^R$ as the Beveridge-Nelson (stochastic) trend of the differential inflation process,

$$\bar{\pi}_t^R = \lim_{j \rightarrow \infty} \mathbb{E}_t (\hat{\pi}_{t+j}^R). \quad (81)$$

The differential inflation rate $\hat{\pi}_t^R$ in this model fluctuates around a stochastic trend given by the aggregate central bank's inflation target. Hence, since we assume in (80) that the target is a random walk, it follows that $\mathbb{E}_t (\hat{\pi}_{t+j}^R) = \tilde{\pi}_t^R$ at any period $j > 0$. In that case, it results from the definition in (81) that $\bar{\pi}_t^R = \tilde{\pi}_t^R$ at every point in time and this confirms that trend and target inflation must be equal in equilibrium also for the differential economy.

Using the differential monetary policy rule in (79) to replace \hat{i}_t^R in (77) – (78), the system of equations that determines the inflation differential and slack differential can be written in the following form,

$$\hat{z}_t^R = A^R \mathbb{E}_t (\hat{z}_{t+1}^R) + a^R (\hat{r}_t^R - \hat{v}_t^R), \quad (82)$$

where,

$$\hat{z}_t^R \equiv \begin{bmatrix} \hat{\pi}_t^R - \bar{\pi}_t^R \\ \hat{x}_t^R \end{bmatrix}, \quad (83)$$

where A^R is a 2×2 matrix and a^R is a 2×1 matrix of structural coefficients. We assume that the process for the aggregate central bank's predicted real rate \hat{v}_t^R is stochastic and exogenous. Under the assumption that the interest rate gap differential $(\hat{r}_t^R - \hat{v}_t^R)$ is stationary, then the system in (55) has a unique nonexplosive

solution in which both \hat{x}_t^R and $\hat{\pi}_t^R - \bar{\pi}_t^R$ are stationary whenever both eigenvalues of the matrix A^R are inside the unit circle. A variant of the Taylor principle which requires that $\psi_\pi + \left(\frac{1-\beta}{k^R}\right) \psi_x > 1$ suffices to ensure the uniqueness and existence of the nonexplosive solution for the differential aggregates. Assuming this condition is satisfied, the solution can be characterized as follows,

$$\begin{pmatrix} \hat{\pi}_t^R \\ \hat{x}_t^R \end{pmatrix} = \begin{pmatrix} \bar{\pi}_t^R \\ 0 \end{pmatrix} + \sum_{j=0}^{\infty} (A^R)^j a^R \mathbb{E}_t \left(\hat{r}_{t+j}^R - \hat{v}_{t+j}^R \right). \quad (84)$$

Hence, the inflation differential is determined by the inflation target differential across both countries and by current and expected future discrepancies between the natural rate of interest differential and the differential of the central bank's own target for the natural rate.

We assume that the central banks adjust their policy rule to track changes in the natural rate of interest that are forecastable one period in advance implying for the differential that,

$$\hat{v}_t^R = \mathbb{E}_{t-1} \left(\hat{r}_t^R \right). \quad (85)$$

Alternatively, we can simply assume—as most of the literature implicitly does—that $\hat{v}_t^R = \hat{r}_t^R + \hat{\varepsilon}_t^{v*}$, where \hat{r}_t^R corresponds to the natural interest rate differential and $\hat{\varepsilon}_t^{v*}$ is an i.i.d. disturbance that captures non-persistent and unanticipated shocks to monetary policy. In either case, the interest rate gap differential $(\hat{r}_t^R - \hat{v}_t^R)$ is viewed as white noise and the solution to the differential system in (82) becomes,

$$\hat{\pi}_t^R = \bar{\pi}_t^R + \lambda^R (\hat{r}_t^R - \hat{v}_t^R) = \bar{\pi}_t^R - \lambda^R \hat{\varepsilon}_t^{v*}, \quad (86)$$

$$\hat{x}_t^R = \mu^R (\hat{r}_t^R - \hat{v}_t^R) = -\mu^R \hat{\varepsilon}_t^{v*}, \quad (87)$$

where the composite coefficients λ^R and μ^R depend on the deep structural parameters of the model. If inflation differential evolve as predicted by this solution, then optimal forecasts of future differential inflation at any horizon $j \geq 1$ must be given by,

$$\mathbb{E}_t \left(\hat{\pi}_{t+j}^R \right) = \bar{\pi}_t^R = \hat{\pi}_t^R - \frac{\lambda^R}{\mu^R} \hat{x}_t^R, \quad (88)$$

or, simply re-arranging, by,

$$\mathbb{E}_t \left(\hat{\pi}_{t+j}^R - \hat{\pi}_t^R \right) = -\frac{\lambda^R}{\mu^R} \hat{x}_t^R. \quad (89)$$

More generally, using the fact that $\hat{\pi}_{t+h|t}^{R,h} \approx \frac{400}{h} \sum_{j=1}^h \hat{\pi}_{t+j}^R$, we can write the forecast h -periods ahead as follows,

$$\mathbb{E}_t \left(\hat{\pi}_{t+h|t}^{R,h} \right) = \frac{400}{h} \sum_{j=1}^h \mathbb{E}_t \left(\hat{\pi}_{t+j}^R \right) = \frac{400}{h} \sum_{j=1}^h \mathbb{E}_t \left(\bar{\pi}_{t+j}^R \right) = 400 \left(\hat{\pi}_t^R - \frac{\lambda^R}{\mu^R} \hat{x}_t^R \right). \quad (90)$$

This implies that no other variable should improve our forecast of changes in the differential inflation if differential slack and the current inflation differential rate are included in the forecasting model. This feature is noted in Woodford (2008) as well and we use it as our key identifying restriction in order to

construct a reduced-form specification (an ADL model) for forecasting inflation that is consistent with the NKPC.

Forecasting future differential inflation using the differential output gap alone would not be accurate since differential inflation potentially has a stochastic trend while differential slack is stationary; one needs to include among the regressors some variable with a similar stochastic trend to that of inflation. But this need not be money growth; current differential inflation itself has the same stochastic trend, so including it to forecast future differential inflation takes care of the trend component without the need to include any other regressors to attempt to track the stochastic trend.

What we need apart from current differential inflation is additional regressors that are stationary and highly correlated with the current deviations of the inflation differential from its stochastic trend. In theory, the differential output gap is one such stationary variable with that property. More generally, what matters is which variables are most useful for tracking relatively high-frequency (or cyclical) variations in inflation differentials. This is true regardless of the horizon over which one wishes to forecast the inflation differential. In this sense, we find that terms of trade or the real exchange rate can be a relevant variable to help us forecast the inflation differential:

Proposition 3 *Domestic terms of trade gap $(\widehat{tot}_t - \widehat{tot}_t)$ and real exchange rate gap $(\widehat{rs}_t - \widehat{rs}_t)$ are proportional to relative slack,*

$$(\widehat{tot}_t - \widehat{tot}_t) \approx \kappa^{tot} \widehat{x}_t^R, \quad (91)$$

$$(\widehat{rs}_t - \widehat{rs}_t) \approx \kappa^{rs} \widehat{x}_t^R, \quad (92)$$

where $\kappa^{tot} \equiv \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right]$ and $\kappa^{rs} \equiv (1 - 2\tilde{\xi}) \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right]$.

Proof. Taking the equilibrium conditions that characterize the terms of trade and the real exchange rate, i.e.

$$\widehat{tot}_t \approx \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right] (\widehat{y}_t - \widehat{y}_t^*), \quad (93)$$

$$\widehat{rs}_t \approx (1 - 2\tilde{\xi}) \widehat{tot}_t = (1 - 2\tilde{\xi}) \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right] (\widehat{y}_t - \widehat{y}_t^*), \quad (94)$$

and re-defining them in terms of deviations from their corresponding potential levels absent any nominal

rigidities, we obtain that,

$$\begin{aligned} \left(\widehat{tot}_t - \widehat{\bar{tot}}_t \right) &\approx \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right] \left(\left(\widehat{y}_t - \widehat{\bar{y}}_t \right) - \left(\widehat{y}_t^* - \widehat{\bar{y}}_t^* \right) \right) \\ &\approx \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right] \widehat{x}_t^R, \end{aligned} \quad (95)$$

$$\begin{aligned} \left(\widehat{rs}_t - \widehat{\bar{rs}}_t \right) &\approx (1 - 2\tilde{\xi}) \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right] \left(\left(\widehat{y}_t - \widehat{\bar{y}}_t \right) - \left(\widehat{y}_t^* - \widehat{\bar{y}}_t^* \right) \right) \\ &\approx (1 - 2\tilde{\xi}) \left[\frac{\gamma}{\sigma\gamma - (\sigma\gamma - 1)(1 - 2\tilde{\xi})^2} \right] \widehat{x}_t^R. \end{aligned} \quad (96)$$

■

In other words, the terms of trade and real exchange rate gap ought to be proportional to the slack differential and, therefore, can be used for forecasting inflation differentials—a fact already noted in Martínez-García and Wynne (2010).

C Historical Estimates of the Model's Parameters

In this section, we provide an extensive analysis on the evolution of the estimates of the model's parameters as mentioned in Section 3.3. In order to obtain historical estimates of the parameters, we conduct 20-year rolling window regressions on the processes underlying the behavior of the Solow residual, the Taylor rule and the monetary policy shock, as well as the import shares and the price elasticity of trade. These estimates are based on U.S. data alone and abstract from some aspects that are essential in the model due to the lack of comparable data—e.g. in the case of estimating spillovers and covariances for the Solow residual. We focus on two periods: 2004:Q1-2008:Q2 for our baseline analysis and 1973:Q4-1983:Q4 for our counterfactuals. The choice of the starting and ending dates of these periods are determined by the observed structural breaks as well as limitations on data. The estimation procedure and results are given as follows.

C.1 Productivity Shock Process Parameters

The estimates are based on the Solow residual, which is calculated as in Martínez-García (2014). The Solow residual reflects the private total factor productivity, excluding government and it is computed as the residual not accounted for with hours worked and capital. Our model excludes government as well, but does not model capital explicitly. Figure C1 shows that the measured productivity persistence is high (top panel). The initial period is relatively lower, but we do not put too much emphasis on it because it could be contaminated by the errors in the pre-1954 period. The volatility (bottom panel) is the parameter for which we analyze in our simulated experiments. There is a very significant decline in the volatility of the productivity shock. It goes from around 1.5 at the beginning to half that amount since 1996. We drop cross-country spillovers and calibrate the covariance between the innovations at the value conventionally assigned in the literature. The literature on Great Moderation also provides important empirical findings on the evolution of these variables over time. The Great Moderation era is mainly characterized by reductions in the conditional variance in time-series models. The variance reduction is generally attributed to a smaller error variance, not to changes in the autoregressive coefficients, as documented by Stock and Watson (2003b), Ahmed et al. (2004), Blanchard and Simon (2001) and McConnell and Pérez-Quirós (2000), among others. Stock and Watson (2003b) calculated a sharp decline in the volatility of the U.S. GDP growth in the first quarter of 1984. Volatility is highest in the 1970s and considerably high in the 1960s and early 1980s.³¹ Moreover, it is also documented that the moderation is not limited to the U.S. Stock and Watson (2003b), for instance, showed evidence from G7 countries.

C.2 Policy Rule Parameters

Since each estimate at a point in time uses data of the previous 20 years, the baseline period estimates use data starting with the Volcker era. The period of Volcker's Policy of Targeting Monetary Aggregates (October 1979 - October 1982) can be excluded from the sample (as in Coibion and Gorodnichenko (2011) and Carlstrom et al. (2009)), but we do not see much difference in the results whether we include that period or not, and therefore include this period. The Taylor rule is computed assuming that the policy rate is the

³¹Stock and Watson (2003b) report the standard deviations of the four-quarter growth rate of real GDP. The standard deviation in the post-1984 period is 0.59 times that of the pre-1984 period. (Standard deviation in the 1970-1980 period is highest but still comparable to its 1960-1970 level.) They also calculate similar volatility declines in macroeconomic variables, including nominal variables such as inflation (GDP deflator) and 90-day T-bill rate.

effective Federal Funds rate, as customarily done. However, the Federal Funds rate starts on 1954Q3, which means we have a slightly shorter sample in the initial estimates reported. Often, researchers concentrate on the post-Korean War period because the data prior to that is subject to many of the distortions remaining from World War II, so we do not believe that it would be an issue to simply ignore the first 6 years in all our structural estimates.

A major problem with the estimation of the Taylor rule, and more generally with our simulated results, is that the model is not well-suited to account for nonlinearities such as setting monetary policy at the zero-lower bound. While we keep the data up to 2011 as in our empirical forecasts, we abstract from the period after 2008 for the evaluation of the forecasting performance and for the analysis of our model—as this is an unusual episode unlike any other in the post-Korean War era.

For consistency with the model and our forecasting exercise, we look at quarter-over-quarter, annualized inflation rates on PCE and the CBO measure of the output gap in my estimates of the Taylor rule.³² The advantage of using the CBO data is that it gives us the longest possible time-series without having to estimate the trend directly, while at the same time giving us quite robust results. The downside is that the CBO measure does not reflect private output only but is total output including the contribution of government. In terms of the inflation measures, using PCE, CPI or the GDP deflator gives similar results. We use the PCE because it is more consistent with the way we define the relative prices in the import equations.

To be consistent with the policy rule in the model, we adopt the same specification as in equation (4), with the assumption of no serial correlation of the monetary policy shocks, $\delta_v = 0$. We abstract from a richer specification of the policy rule to keep our analysis as tractable as possible while we obtain similar qualitative results with the literature.

Figure C2 below depicts the historical estimates of the Taylor rule parameters which (i) include inflation response, Ψ_π (top left), (ii) output gap response, Ψ_x (top right), (iii) the volatility of the monetary policy shock, σ_v (bottom left), and (iv) the implied probability of violating the Taylor rule (bottom right). It provides evidence on the instability of the point estimates for the responses to inflation and output gap, especially on the response to inflation. The implied probability of violating the Taylor principle is positive until the late 1990s with the inflation and output gap responses moving into the indeterminacy region. A remarkable result is on the size of the volatility of the monetary shock: it has changed over this period going from below 0.5 to more than 6 times greater between 1968 and 1980, staying above 2 through mid-1990s, and then abruptly declining afterwards. This proves a strong case for variation in the volatility that needs to be explored.

Coibion and Gorodnichenko (2011), among others,³³ also provide historical estimates of the coefficients of a generalized Taylor rule.³⁴ Their estimates of Ψ_x do not show much variation from the late 1960s to the early 2000s. They indicate that both the inertia of the monetary policy and the parameter on inflation gap have increased recently. Their time varying estimate for Ψ_π is relatively high in the late 1960s, as well as the early 1980s and onwards, but low during the 1970s. Rudebusch (2006) provides evidence that for the 1990s, a positive inertia parameter in the policy rule is more plausible. However, a noninertial policy rule

³²Interestingly, the year-over-year inflation rates offer a more stark representation of the Taylor rule with a very consistent estimate for the inflation parameter until 2008, however, the metric is somewhat inconsistent with those used in the forecasts.

³³For other historical estimates of Taylor Rule coefficients for the U.S., see also Judd and Rudebusch (1998), Taylor (1999) and Clarida et al. (1999).

³⁴The policy rule includes time-varying parameters (with a response to output growth) as well as time-varying trend inflation and interest smoothing of order two.

is a common benchmark in the literature (e.g. Taylor (1993) and Feldstein et al. (2004)) and is therefore a natural case to investigate.

Stock and Watson (2003b) provide a helpful comparison of the monetary policy shock volatilities for the pre-1983 and post-1984 era. Using structural VAR and implementing the methodologies of Christiano et al. (1997) and Bernanke and Mihov (1998),³⁵ they compute the implied monetary policy innovations. Volatility of these shocks exhibited a decline in the Great Moderation era, following high volatility in 1960 – 83 and having a peak during 1979 – 83.³⁶ Similarly, Smets and Wouters (2007), report a decline in the volatility of monetary policy shocks (as well as productivity shocks) in the U.S. in an extended DSGE model for the Great Moderation era (1984 – 2004) relative to the Great Inflation era (1966 – 79), consistent with the evolution of our estimates.

C.3 Trade Parameters

We estimate an import equation that would be consistent with virtually any IRBC or open-economy NK model that we can write which features CES aggregation of foreign and domestic goods, including the one in Martínez-García and Wynne (2010). The import demand relationship up to a first-order approximation is given in log deviations from steady state by,

$$imp_t \approx -sigma * (P_t - P_t^{imp}) + (cons_t + inv_t)$$

where imp_t represents the demand for real imports, the estimated elasticity parameter is $-sigma$ (where $sigma$ is the structural parameter of our model), P_t represents the price index for domestic absorption, P_t^{imp} is the price index corresponding to the imported goods, and $cons_t + inv_t$ represents the domestic absorption. In order to estimate this regression we assume a linear trend as well, which is equivalent to detrending the data prior to the estimation. Since we do not have enough detailed data to distinguish between imports by the private and public sector, we assume that the elasticity of substitution would be the same for both; hence we incorporate government consumption and investment in the measure of absorption. However, we do have data for a sub-period on imports of goods excluding oil and the literature has documented that the price elasticity for oil is significantly lower than for other traded goods. We take advantage of the data available and report estimates for both total imports and non-oil imports of goods (and in the regression use their corresponding deflators).

Values traditionally reported for the estimation of sigma range from the classic IRBC value of 1.5 for the elasticity of substitution between domestic and foreign goods (Backus et al. (1994)) to as high as 6 (Anderson and van Wincoop (2004), based on the trade literature). Our findings seem consistent for most of the sample with the estimates in the lower end of that range, around 1.5. Figure C3 depicts results for the openness parameter estimates. Accordingly, total imports have gone up over time from around 5% at the beginning of the sample to around 15% at the end. If we exclude services and oil, the share drops to less than 10% by the end of the sample. Therefore, the 6% that we had originally selected in our calibration seems consistent with this finding. Also, the relevant trade elasticities should be the ones that exclude oil to be consistent with the model. In that case, we see that for most of the sample we cannot reject the null

³⁵They take into account that the monetary policy shifted over the sample period.

³⁶Volatility of money shocks during 1984-2001 is about 0.50 times the volatility in 1960-1983 and about 0.76 times the volatility in 1960-1978 period according to CEE methodology.

hypothesis that they are equal to 1, which is a little lower than our original calibration. This means that a Cobb-Douglas aggregator, rather than a CES aggregator, would work in this situation. There is an anomaly at the end of the sample that can easily be due to the effects of the financial recession.

The fact that there are no data for us to identify the same structural parameters in the rest of the world means that we are very limited in the kinds of experiments that we can realistically conduct. Still, we believe we make an informed evaluation of the mechanics of the model that would be consistent with standard practice and would help us gain insight on the forecasting performance of the variables of interest.

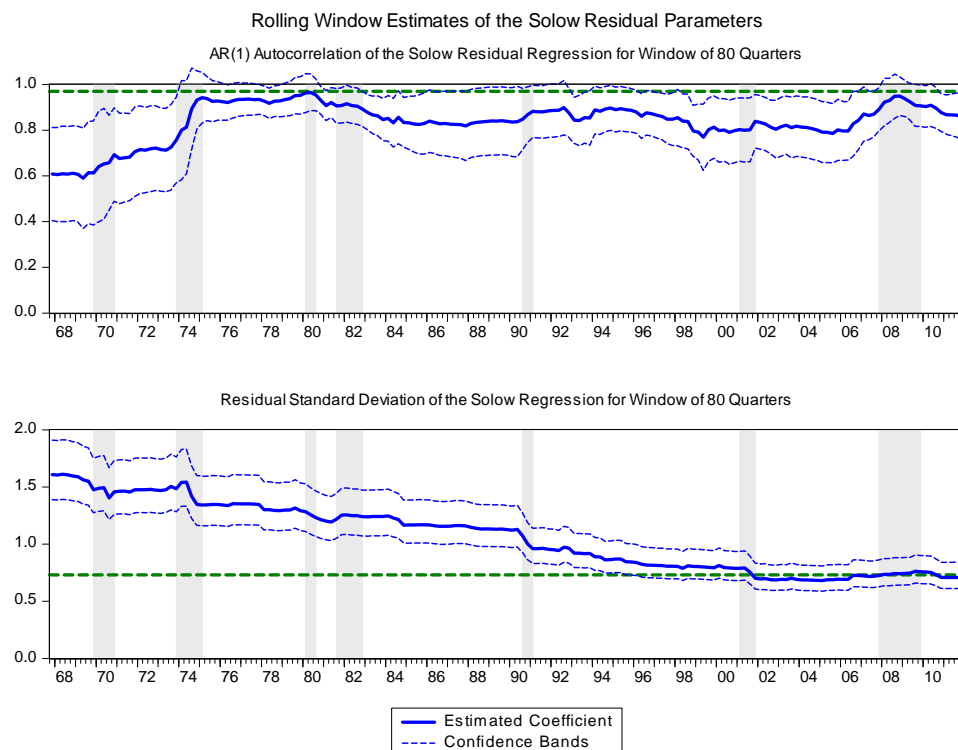


FIGURE C1. Evolution of the Solow residual parameters

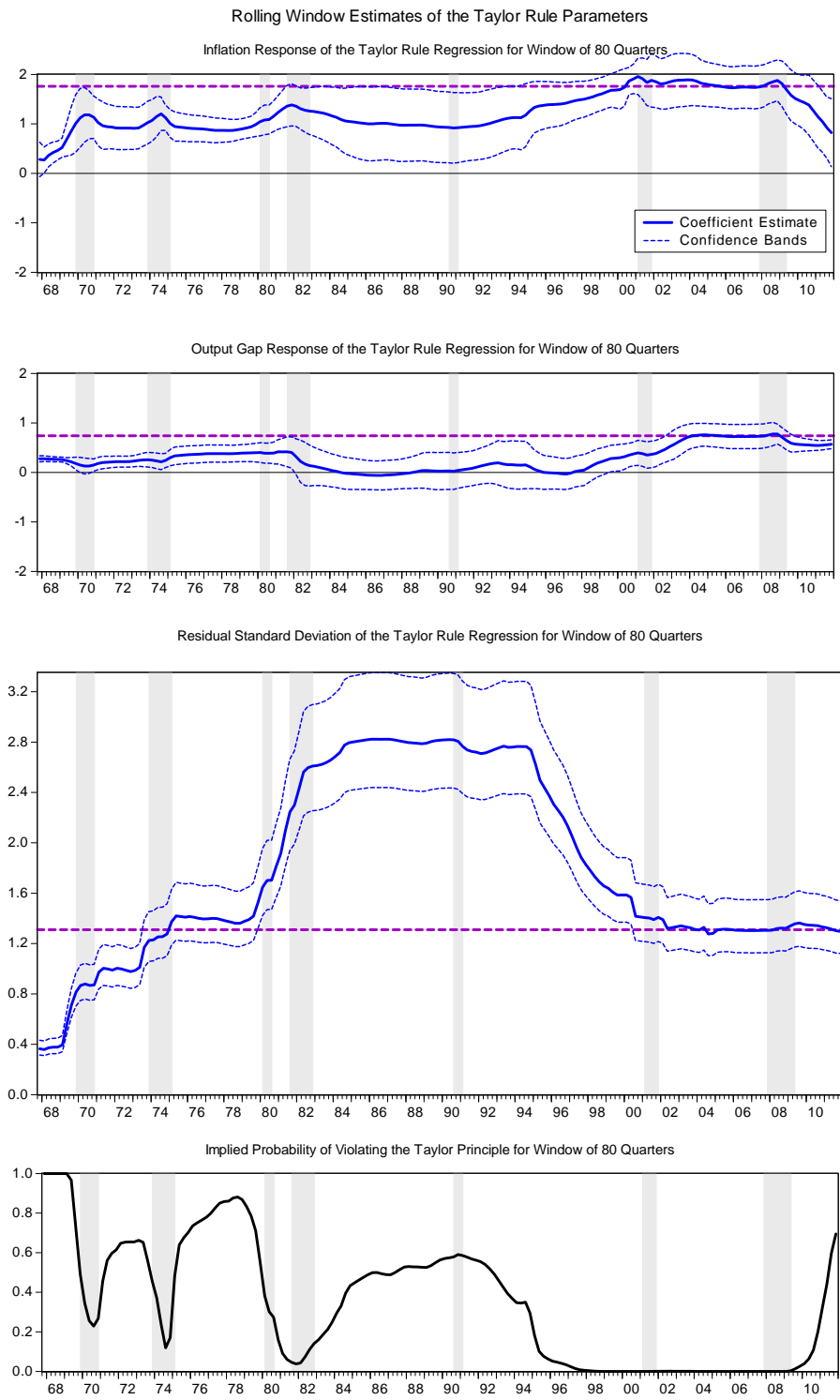


FIGURE C2. Evolution of the Taylor rule parameters

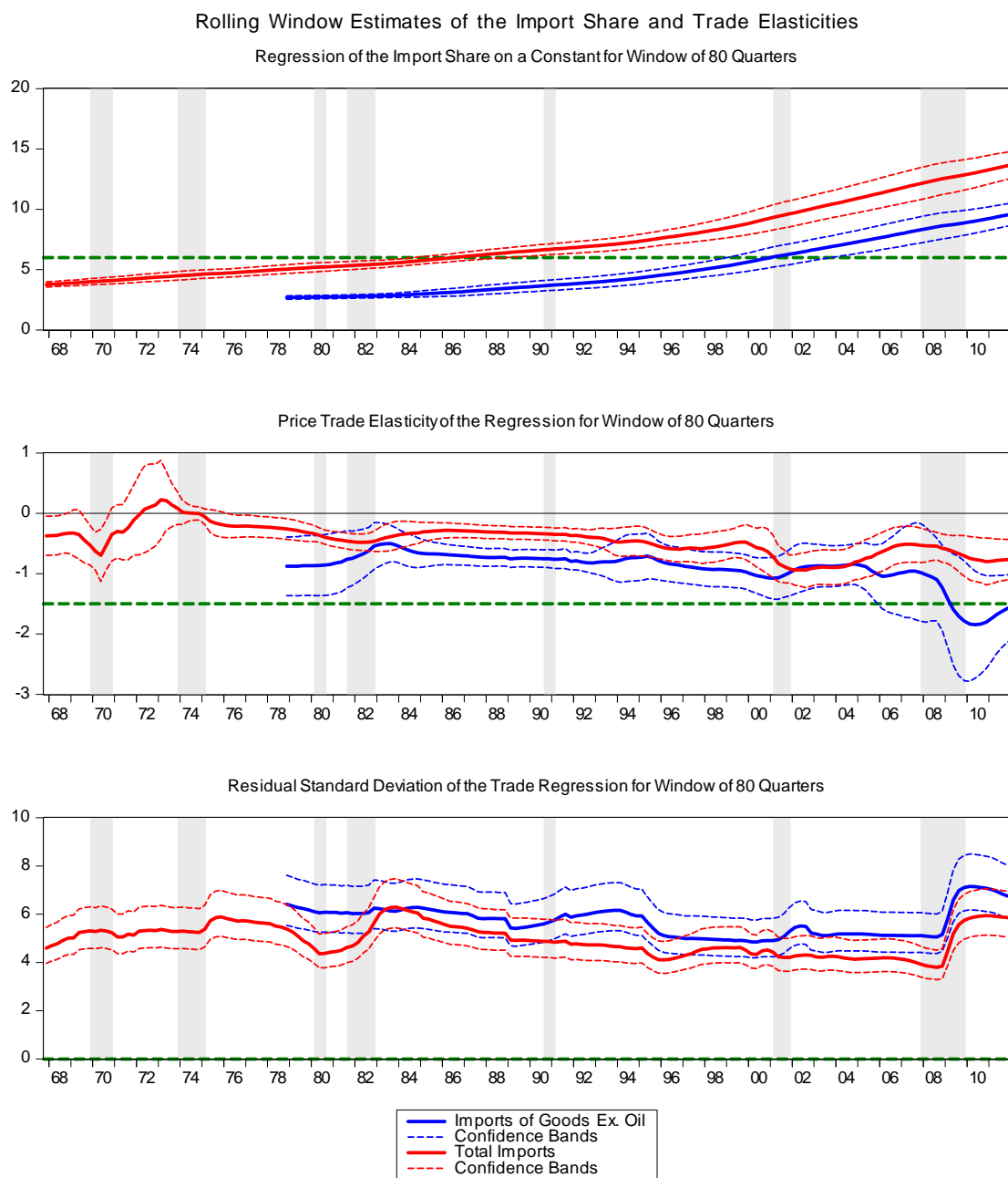


FIGURE C3. Evolution of the openness parameters

D Data Description

This section gives details for the data used in the empirical forecasts only.

Abbreviations

BEA = U.S. Bureau of Economic Analysis; BLS = U.S. Bureau of Labor Statistics; BBK = German Federal Bank; BIS = Bank for International Settlements; CAO = Cabinet Office (Japan); CBI = Confederation for British Industries; CBO = Congressional Budget Office; FRB = Federal Reserve Board; FRBD = Federal Reserve Bank of Dallas; FRED = Federal Reserve Economic Data (St. Louis Fed); IMF = International Monetary Fund; INSEE = National Institute of Statistics and Economic Studies (France); ISTAT = Istituto Nazionale Di Statistica (Italy); OECD = Organisation for Economic Cooperation and Development; OECDMEI = OECD Main Economic Indicators; ONS = Office for National Statistics (UK); SAAR = Seasonally adjusted at an annual rate; SA = Seasonally adjusted; SCAN = Statistics Canada; WK = Wright Killen & Company; WSJ = Wall Street Journal.

All series are quarterly unless otherwise indicated and obtained from Haver Analytics. In general, we indicate the original source if the series is available outside Haver Analytics. While we try to be consistent in terms of the definitions across countries, under cases in which data availability is limited, we use the series with the closest definition.

1. Price series

Series used for U.S. inflation: All series are seasonally adjusted. Start dates of the series vary across different measures and they all end in 2011:Q4. Base years are set such that 2005:Q1-2005:Q4=100. Start dates of each series are indicated in parentheses. We take CPI (all items) (1947:Q1) from the BLS, core CPI (all items ex. food and energy) (1957:Q1) from the BLS, GDP implicit price deflator (1947:Q1) from the BEA; PCE chain price index (1947:Q1) from the BEA, Core PCE (all items ex. food and energy) chain price index (1977:Q1) from the BEA and PPI (finished goods) (1947:Q2) from the BLS, Sticky Prices (Sticky CPI) (1967:Q1) from the Atlanta Fed, Sticky Prices ex. shelter (Core sticky CPI) (1967:Q1) from the Atlanta Fed. *Series used for terms of trade:* We use exports and imports under the heading ‘price indexes for GDP’ in National Income and Product Accounts in the BEA to calculate U.S. terms of trade. Both series are seasonally adjusted, with the base year 2005=100 and cover periods 1947:Q1-2011:Q4. The terms of trade series is calculated as $100 \times \text{export price index} / \text{import price index}$. The terms of trade ex. oil is calculated using the imports of non-petroleum goods (chain price index) and exports of goods (chain price index) from the BEA (1967:Q2-2011:Q4). *Oil prices:* We use West Texas Intermediate spot oil price (\$/barrel, prior ’82=posted price) from WSJ (1946:Q1) and Saudi Arabian Light Crude spot oil price (\$/barrel) from WK.

We apply two filtering (detrending) techniques on terms of trade, terms of trade ex. oil, WTI and SAL prices: a one-sided Hodrick-Prescott (HP) filter and first differencing. The HP filter is applied as described in Stock and Watson (1999b). This is a one-sided HP filter derived using the Kalman filter to optimally filter the series that renders the standard two-sided filter optimal. The first differenced terms of trade is calculated as $\Delta ToT_t = ToT_t - ToT_{t-1}$, and similarly for all other series.

2. Monetary aggregates

All series are seasonally adjusted and quarterly (end-of-period aggregates of monthly series). Our U.S. series are from FRED, for 1948:Q1-2011:Q4. For UK, we have M4 series available from OECD (1963:Q1-2011:Q4). For other countries, data become limited for certain periods and sources and therefore we splice

two series. Therefore we obtain M3 for Canada from BIS (1962:Q1-1981:Q4) and OECD (1982:Q1- 2011:Q4); M2 for Germany from BIS (1963:Q1-1990:Q4) and BBK (1980:Q1-2011:Q4); two spliced M2 series for Italy from Bank of Italy (1948:Q4-1998:Q4/1997:Q1-2011:Q4); M2 for Japan from Bank of Japan (1963:Q1-1966:Q4) and FRED (1967:Q1-2011:Q4). For France, we splice three series, M2 from NBER (<http://papers.nber.org/books/darb83-1>), M2R, and M3 from BIS (1963:Q1-1969:Q4 and 1970:Q1-2011:Q4, respectively). For France, Germany and Italy, the first part of the series is converted from the national currency to Euros using the European Currency Unit (1999). These series help us construct a G7 money growth series starting in 1955:Q2 by taking the average of money growth rate of G7 countries.

3. Slack measures

All measures used cover the period 1980:Q1-2011:Q4 unless stated otherwise.

CBO U.S. slack: Defined as ‘Output Gap in Percentage of Real GDP’, and is calculated as

$$\frac{100 \times (RPGDP_t - RGDP_t)}{RGDP_t}$$

where $RPGDP_t$ and $RGDP_t$ are real potential GDP and real GDP at quarter t , respectively (SAAR, Billions of Chained 2005 Dollars). We take our real GDP series from BEA and real potential GDP series from CBO. The U.S. HP-filtered series is simply quarterly U.S. real GDP series with HP filter applied. Then the logs of the cyclical component is taken and multiplied by 100.

FRBD U.S. slack: The series is constructed by the FRBD, and the methodology can be described as follows. First, the Phillips curve is estimated with annualized quarterly inflation (specifically, core CPI) and unemployment rate/capacity utilization rate. The regression equation for this is specified as is constructed as follows:

$$\pi_t = \alpha_1 + \alpha_2 \pi_{t-1} + \alpha_3 \pi_{t-2} + \alpha_4 \pi_{t-3} + (1 - \alpha_2 - \alpha_3 - \alpha_4) \pi_{t-4} + \alpha_5 ur_t + \epsilon_t$$

where $\pi_t = 400 \times \log(p_t/p_{t-1})$, p_t is the price index, ur_t is unemployment rate where we define the potential unemployment rate as $ur^* = -\hat{\alpha}_1/\hat{\alpha}_5$. We run a similar regression with the capacity utilization rate, $capu_t$ and define the potential rate of capacity utilization, $capu^* = -\hat{\alpha}_1/\hat{\alpha}_5$, similarly.

Then the slack measure is computed by running the following regression

$$\pi_{t+4} - \pi_t = -\beta_1(ur_t - ur^*) + (1 - \beta_1)(capu_t - capu^*) + \epsilon_t$$

and the slack measure is calculated as $slack_t = -\hat{\beta}_1(ur_t - ur^*) + (1 - \hat{\beta}_1)(capu_t - capu^*)$.

FRBD G7 slack: Produced by the FRBD and calculated by applying the procedure described above for each member of the G7 economies. After obtaining the ‘domestic slack measure for a given country, the GDP shares of each country is calculated so that for country i at quarter t , $share_{i,t} = GDP_{i,t} / \sum_i GDP_{i,t}$. The G7 slack is the GDP-weighted average of the slack measures of individual countries.

The data series we use here are as follows:

- GDP series to construct the GDP shares of each country (sources indicated in parentheses): Canada (SCAN), France (INSEE), Germany (BBK), Italy (ISTAT), Japan (CAO), UK (ONS), U.S. (BEA).

All series are in billions of U.S. Dollars, seasonally adjusted (1978:Q1-2011:Q4). For France, Germany and Italy, the series are working day adjusted.

- Capacity utilization rates for each country (SA, %) are given as follows. We use U.S. Capacity utilization series in manufacturing, 1948:Q1-2011:Q4 from FRB. France: manufacturing capacity utilization rates come from manufacturing surveys, covering the period 1978:Q4-2011:Q4. To extend the series, we use capacity utilization series calculated by Cette (1990) that date back to 1968:Q4. Germany: We use manufacturing surveys from OECDMEI (1968:Q4-2011:Q4). United Kingdom: The UK manufacturing survey series for 1971:Q1-2011:Q4 from OECDMEI is extended back in time using the following normalized series from Datastream (available for 2/15/1950-8/15/2011, originally from CBI UK, % working below capacity). Japan: The OECD series covers 1978:Q1-2011:Q4. We splice the series with Bloomberg data obtained from OECD Japan prod. industry total rate of capacity utilisation IPB 2005 that starts in Feb-1968. Canada: We use series from SCAN which starts in 1987:Q1 splicing together with a historical series with the manufacturing survey capacity utilization rate (NSA, %), starting in 1962:Q1 (OECDMEI).
- As a measure of inflation, we use core CPI. All series are seasonally adjusted, come from OECDMEI and the base year is 2005=100 for all countries with the exception that the base year is 2010=100 for Japan and 1982-84=100 for the U.S..

FRBD G28 Slack: This measure is calculated by HP filtering of FRBD G28 index which uses constant 2005 (PPP adjusted) weights to aggregate GDP series of the 28 countries: Argentina, Australia, Brazil, Canada, Chile, China, Colombia, France, Germany, Hong Kong, India, Indonesia, Ireland, Israel, Italy, Japan, Korea, Malaysia, Mexico, Netherlands, Phillipines, Saudi Arabia, Singapore, Sweden, Switzerland, Thailand, UK, Venezuela and U.S. GDP series used are quarterly; and for some countries for which only disaggregated (annual) data are available, we apply the quadratic match average method to interpolate these series.

IMF U.S. and IMF Advanced Slack: Both slack measures are defined as 'Output Gap in Percentage of Real GDP (%)' for the U.S. and for a group of advanced countries (Australia, Austria, Belgium, Canada, Cyprus, Denmark, Finland, France, Germany, Greece, Ireland, Italy, Japan, Netherlands, New Zealand, Norway, Portugal, Spain, Sweden, UK and U.S.). These measures are published by IMF WEO, annually and available between 1980-2011. Therefore we interpolate the series by the quadratic match average method to disaggregate into quarterly frequency.

OECD U.S., OECD G7 and OECD Total Slack: All three measures are defined as the 'Output Gap of the Total Economy (%)', published by OECD Economic Outlook. OECD Total consists of 30 OECD countries: Australia, Austria, Belgium, Canada, Cyprus, Czech Republic, Denmark, Finland, France, Germany, Greece, Iceland, Ireland, Italy, Japan, Korea, Luxembourg, Mexico, Netherlands, New Zealand, Norway, Portugal, Slovakia, Spain, Sweden, Switzerland, Turkey, UK and U.S. and the series go back to 1970:Q4.

U.S. HP-filtered GDP: Calculated using quarterly U.S. real GDP series from BEA. First, the logs of the series is taken and multiplied by 100 and then Hodrick-Prescott filter is applied.

4. Credit series

We use end-of-period, quarterly, long series on credit to non-financial sectors from the BIS for G7 countries. While data for various borrower-lender combinations are available, we choose lenders from 'all sectors' and borrowers from 'private sector' in order to be able to go back in time as far as possible. The series are adjusted for breaks for all countries except for the U.S. where only unadjusted series are available. For

the UK the series start in 1963:Q1 (Pound Sterling), for Italy in 1960:Q4 (Euro), for the U.S. in 1952:Q1 (U.S. Dollar), for Japan in 1964:Q4 (Japanese Yen), for France 1969:Q4 (Euro), for Germany 1948:Q4 (Euro), and for Canada in 1954:Q1 (Canadian Dollar) and all series are available until 2011:Q4. For the period 1965:Q1-1969:Q4, we use all countries except France to calculate global average credit growth. For 1970:Q1 onwards, we use G7 countries to calculate the average credit growth.

5. REER series

We use the BIS series for U.S. REER (narrow definition) since it is the longest series available. The series covers the 1964:Q1-2011:Q4 period, with the base year set as 2005=100 (average). For our forecasts, we use the first differenced and HP-filtered (1-sided) REER series.

E The Bootstrap Algorithm in Three-variable Forecasts

To forecast h -quarter ahead inflation, $\hat{\pi}_{t+h|t}^h$, we evaluate the predictive ability of pairs of variables in the general form

$$\hat{\pi}_{t+h|t}^h = \alpha_1 + \alpha_{21}(L) \hat{\pi}_t + \alpha_{22}(L) \hat{x}_{1t} + \alpha_{23}(L) \hat{x}_{2t} + \hat{\epsilon}_{t+h}$$

where \hat{x}_{1t} and \hat{x}_{2t} denote either (i) domestic slack and foreign slack, (ii) domestic slack and terms of trade (first differenced), or (iii) domestic money supply growth and foreign money supply growth.

We calculate MSFE-F statistics to test the null hypothesis that the MSFE of the naïve forecast is higher than or equal to the MSFE of the augmented model above. We calculate critical values based on a simple parametric bootstrap algorithm with 5000 replacements.³⁷ In this case, the DGP involves the estimation of a 3-equation VAR and uses the residuals. The first equation is the AR process of inflation, $\hat{\pi}_t$. The remaining two equations are the equations for the predictors where we include the distributed lags of all three variables

$$\hat{\pi}_t = \beta_1 + \beta_2(L) \hat{\pi}_t + \hat{\epsilon}_{1,t}$$

$$\hat{x}_{1t} = \theta_1 + \theta_{11}(L) \hat{\pi}_t + \theta_{12}(L) \hat{x}_{1t} + \theta_{13}(L) \hat{x}_{2t} + \hat{\epsilon}_{2,t}$$

$$\hat{x}_{2t} = \gamma_1 + \gamma_{11}(L) \hat{\pi}_t + \gamma_{12}(L) \hat{x}_{1t} + \gamma_{13}(L) \hat{x}_{2t} + \hat{\epsilon}_{3,t}$$

The lag length is limited to four for each regressor and selected based on SIC.

E.1 Results

³⁷As suggested by Clark and McCracken (2006) in nested models the F-statistics have non-standard, asymptotic distributions and therefore a bootstrap procedure is needed to calculate the empirical critical values.

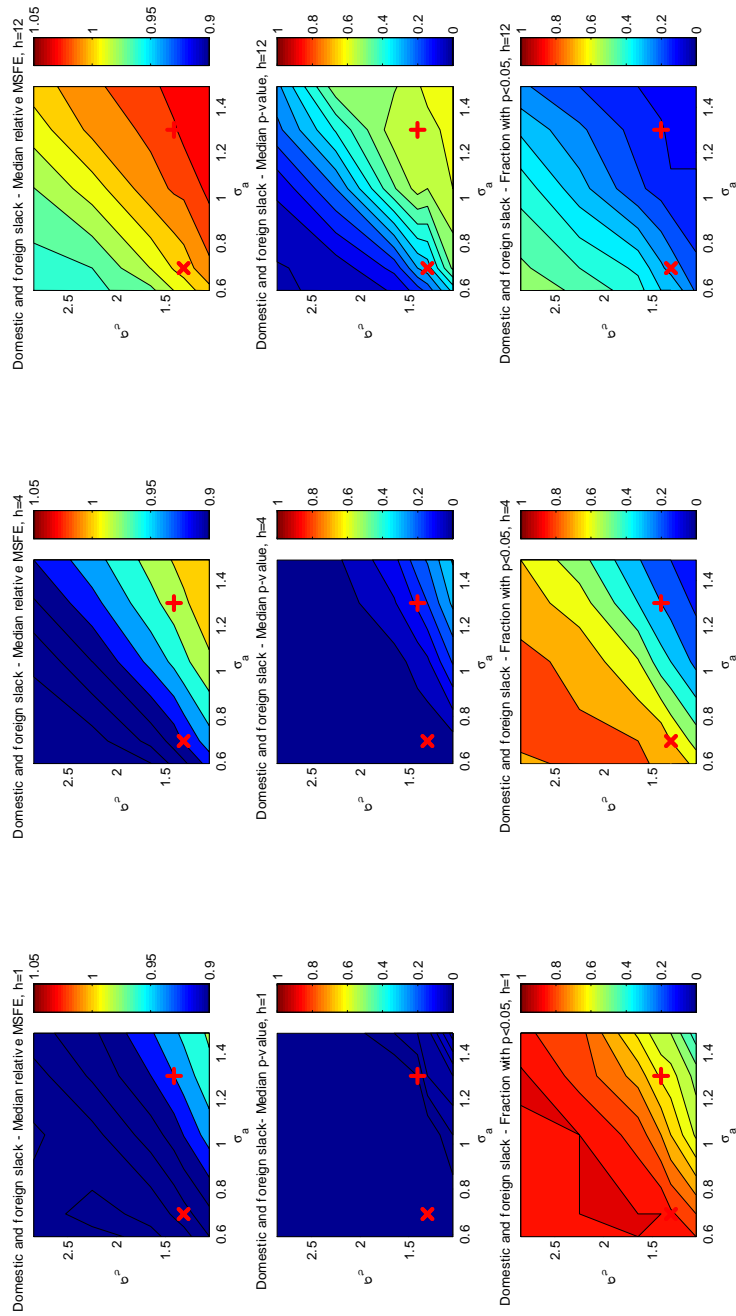


FIGURE E1. Model's prediction of the relative MSFEs of forecasts with domestic and foreign slack as a function of the parameters of *good luck*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).
Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.
Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

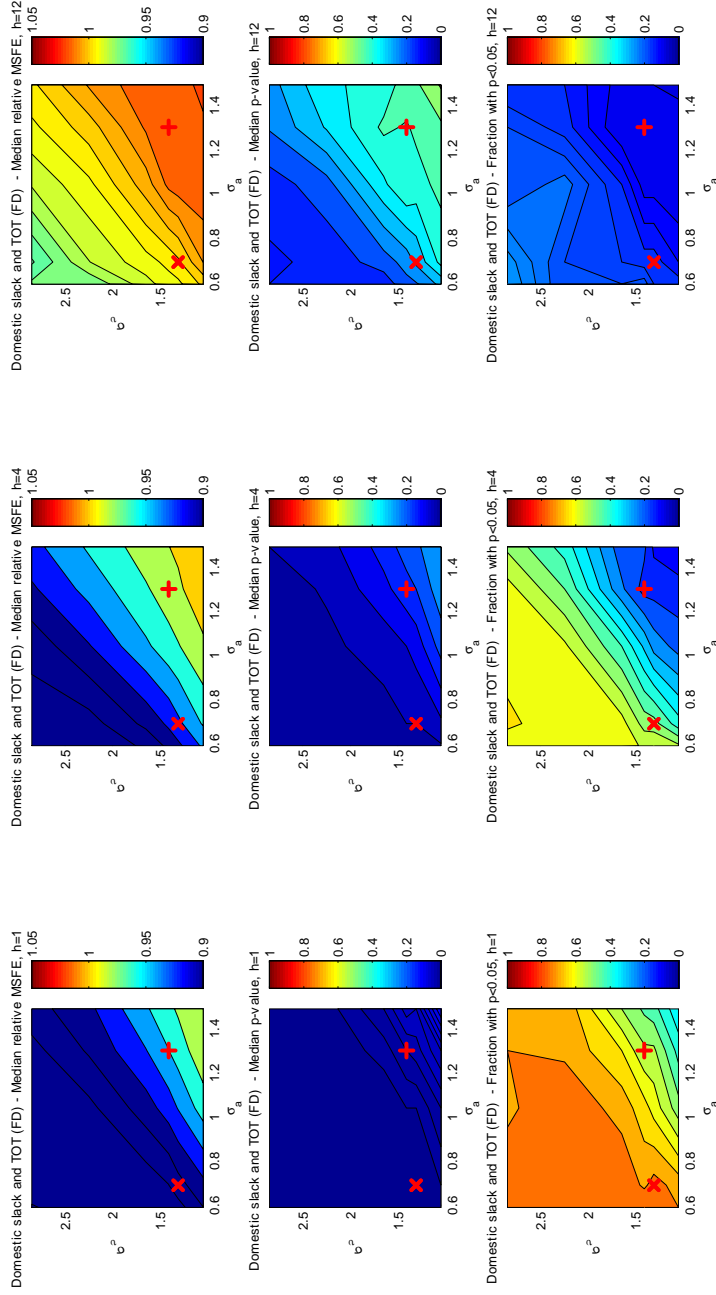


FIGURE E2. Model's prediction of the relative MSFEs of forecasts with domestic slack and TOT (FD) as a function of the parameters of *good luck*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).
Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.
Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

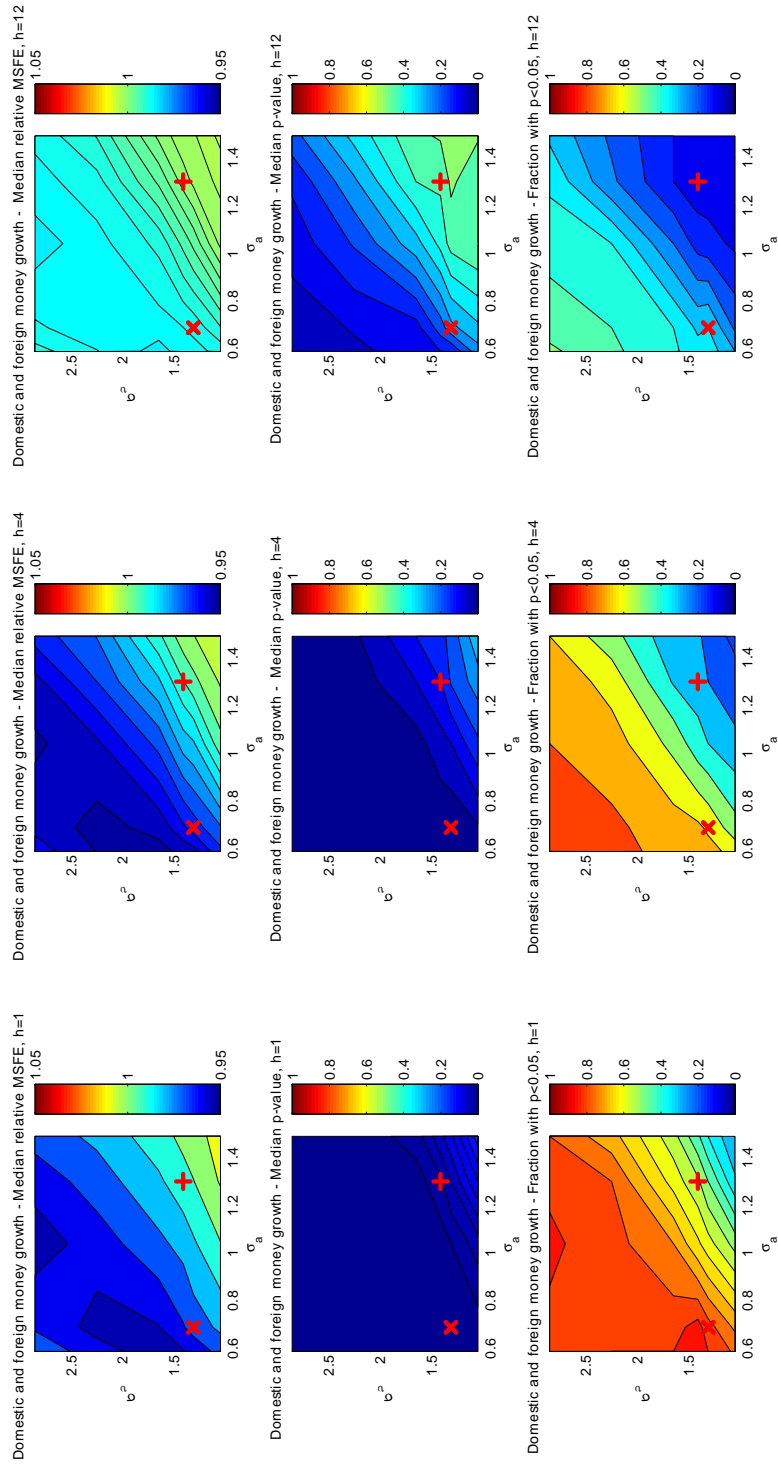


FIGURE E3. Model's prediction of the relative MSFEs of forecasts with domestic and foreign money supply growth as a function of the parameters of *good luck*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).

Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.

Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

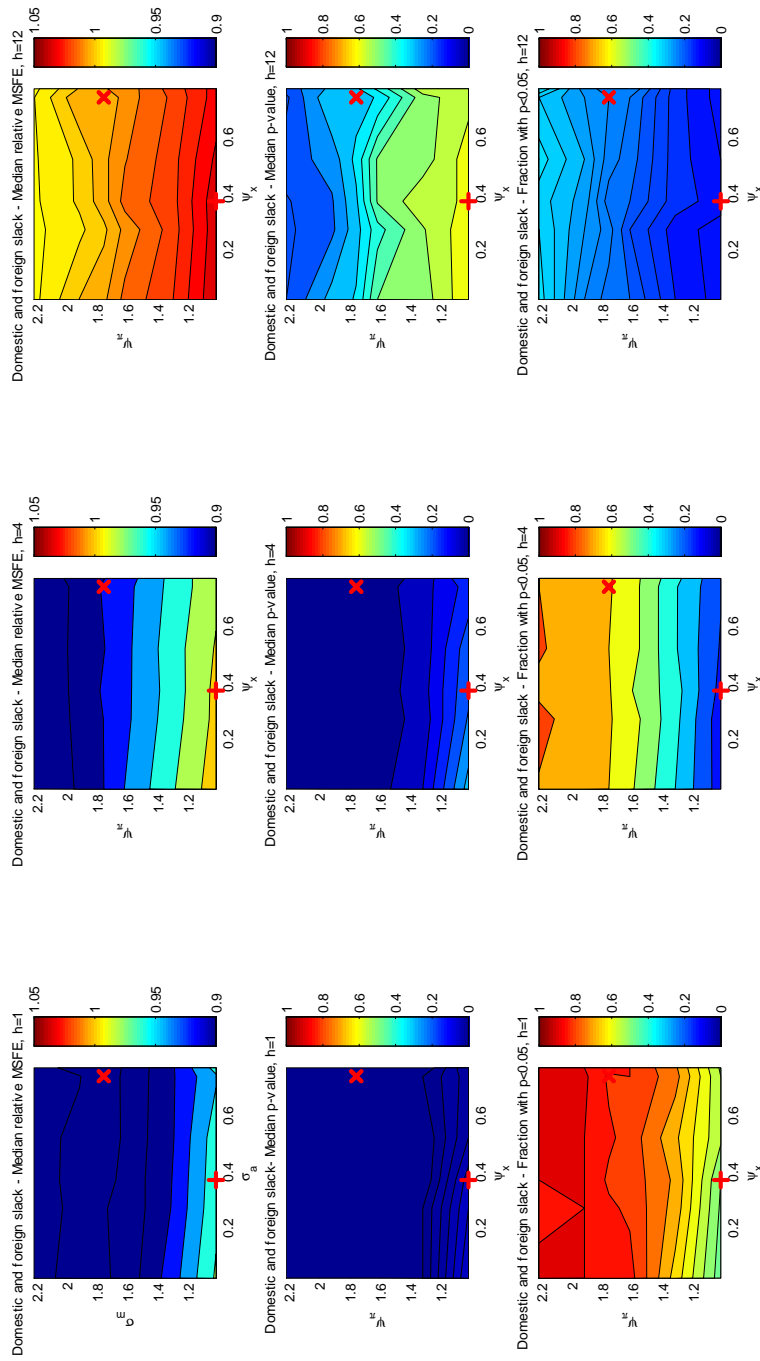


FIGURE E4. Model's prediction of the relative MSFEs of forecasts with domestic and foreign slack as a function of the parameters of *monetary policy*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).

Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported. Period 1 and Period 2 (benchmark) parameters are denoted by ' v ' and ' w ', respectively.

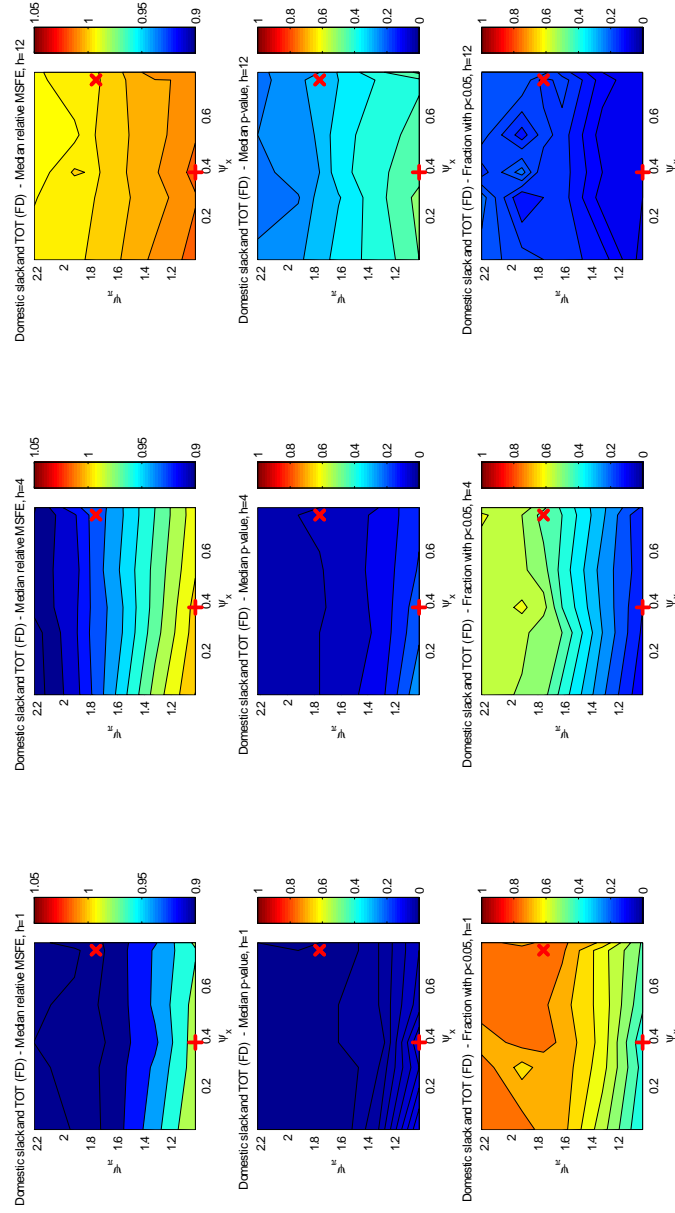


FIGURE E5. Model's prediction of the relative MSFEs of forecasts with domestic slack and TOT (FD) as a function of the parameters of *monetary policy*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).
Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.
Period 1 and Period 2 (benchmark) parameters are denoted by 'x' and 'x', respectively.

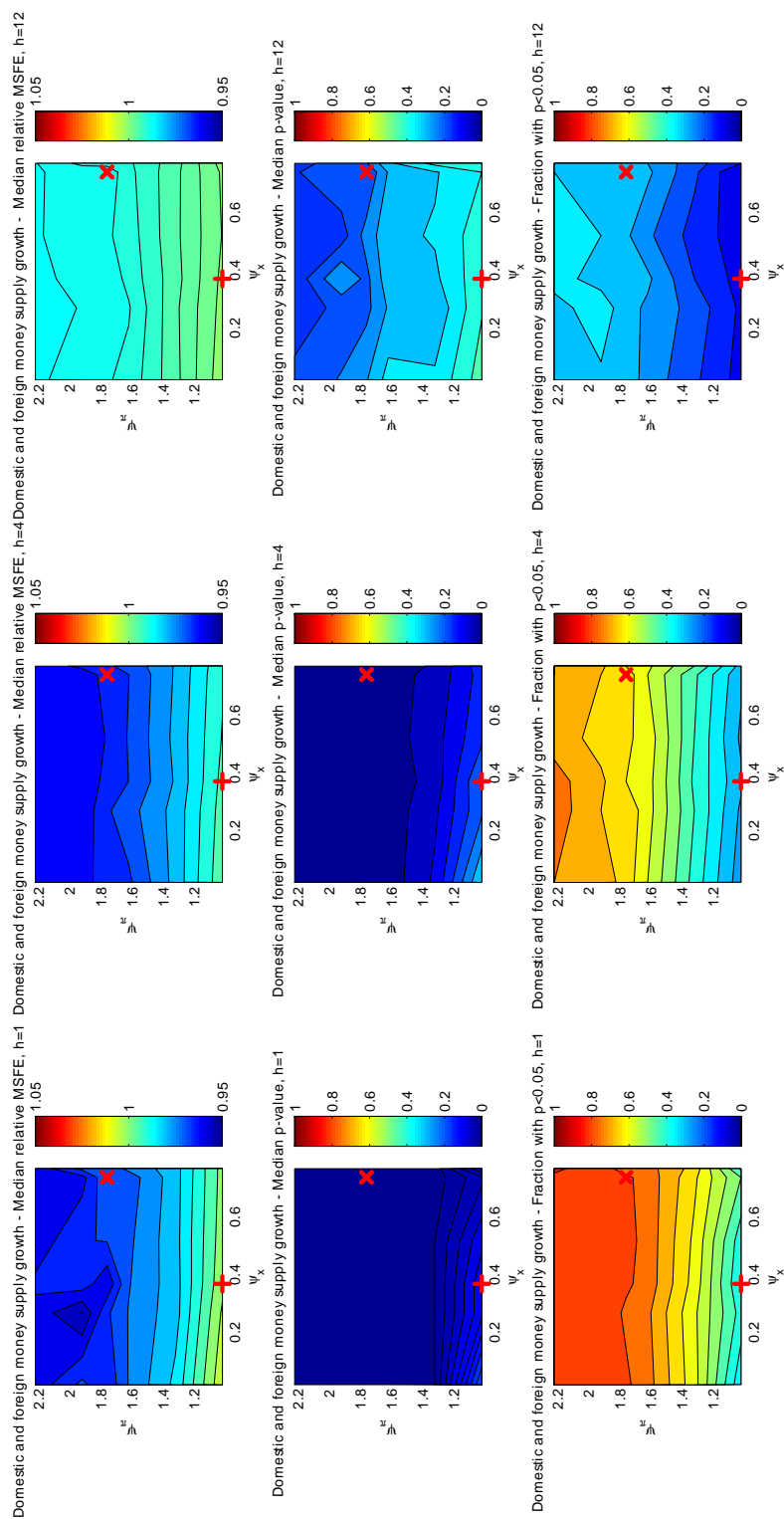


FIGURE E6. Model's prediction of the relative MSFEs of forecasts with domestic and foreign money supply growth as a function of the parameters of *monetary policy*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).
Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.
Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

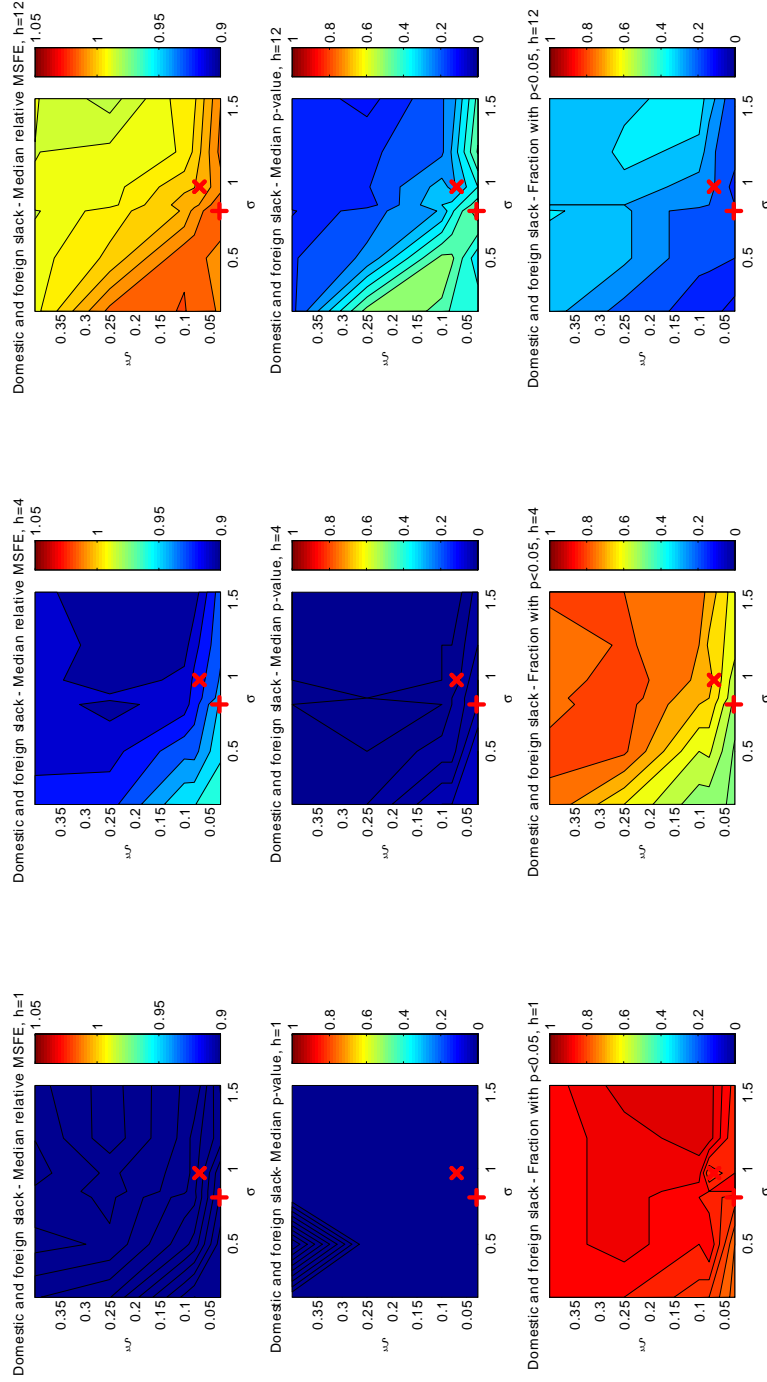


FIGURE E7. Model's prediction of the relative MSFEs of forecasts with domestic and foreign output gap as a function of the parameters of *openmess*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).
 Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.
 Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

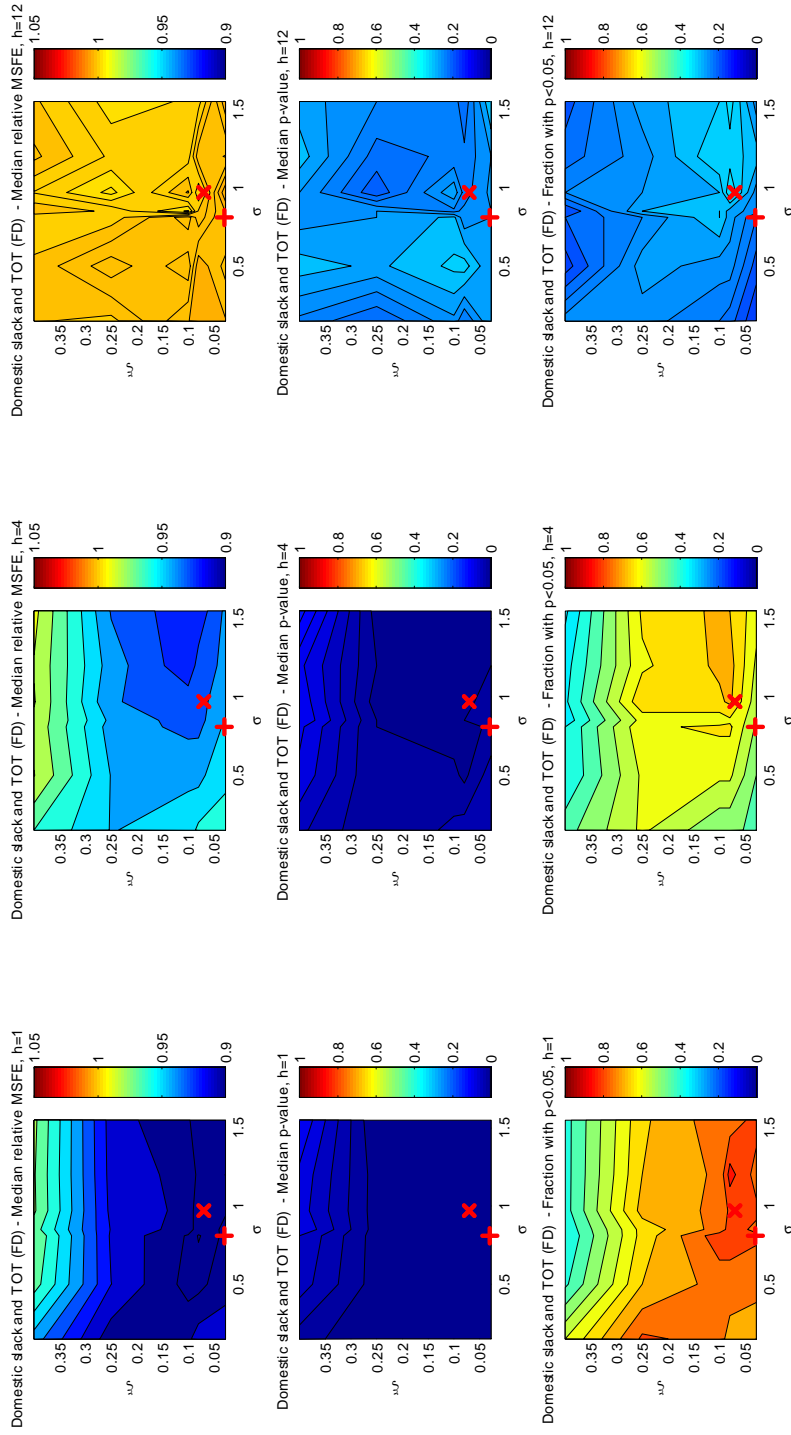


FIGURE E8. Model's prediction of the relative MSFEs of forecasts with domestic slack and TOT (FD) as a function of the parameters of *opemess*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).
 Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.
 Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

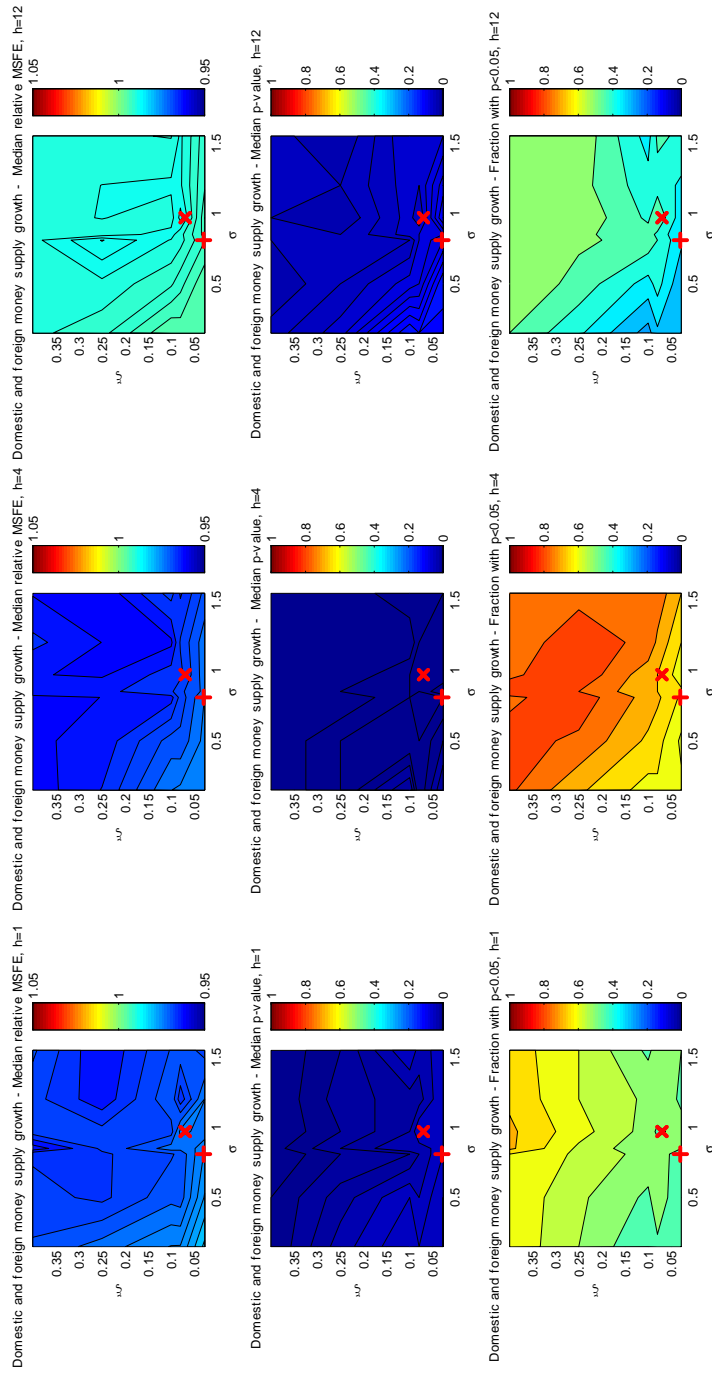


FIGURE E9. Model's prediction of the relative MSFEs of forecasts with domestic and foreign money supply growth as a function of the parameters of *openness*

Note: MSFEs are relative to the MSFEs of the univariate AR process of inflation (restricted model).

Median MSFEs, median p-values and fraction of statistically significant MSFEs in 100 simulations are reported.

Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

E.2 Correlations

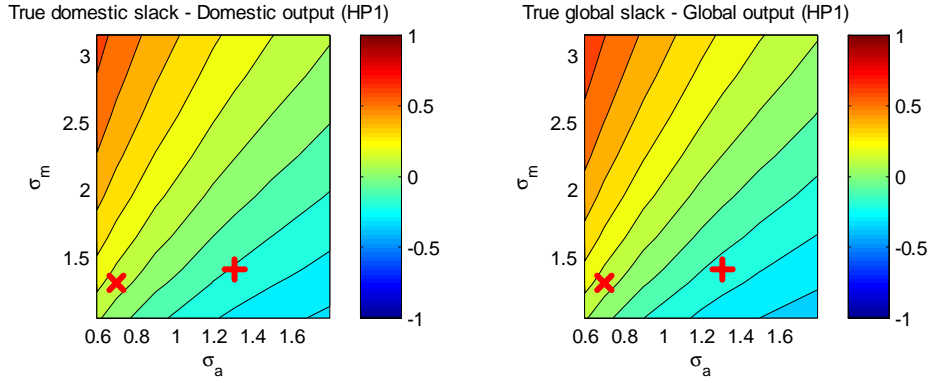


FIGURE E10. Correlations of variables as a function of the parameters of *good luck*
Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

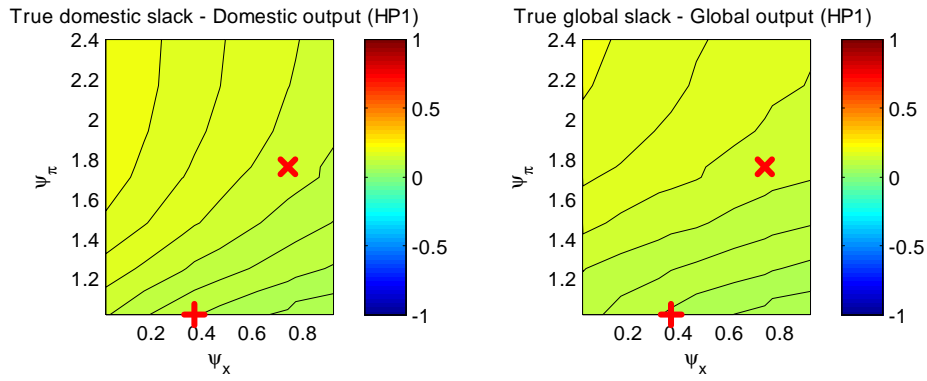


FIGURE E11. Correlations of variables as a function of the parameters of *monetary policy*
Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

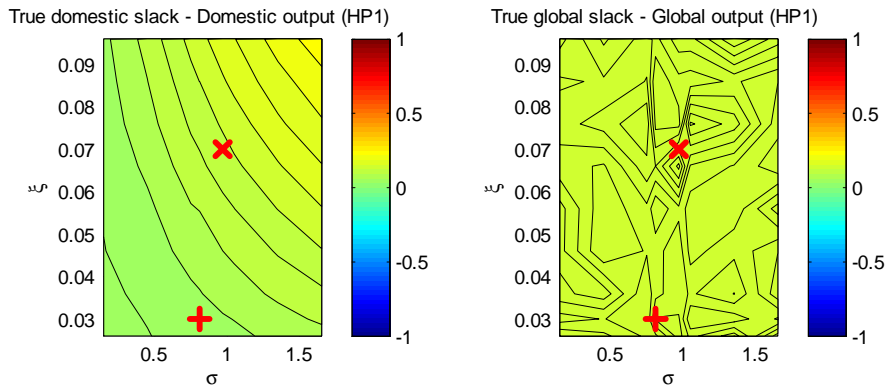


FIGURE E12. Correlations of variables as a function of the parameters of *openness*
Period 1 and Period 2 (benchmark) parameters are denoted by '+' and 'x', respectively.

F Tables and Figures for the Empirical Section

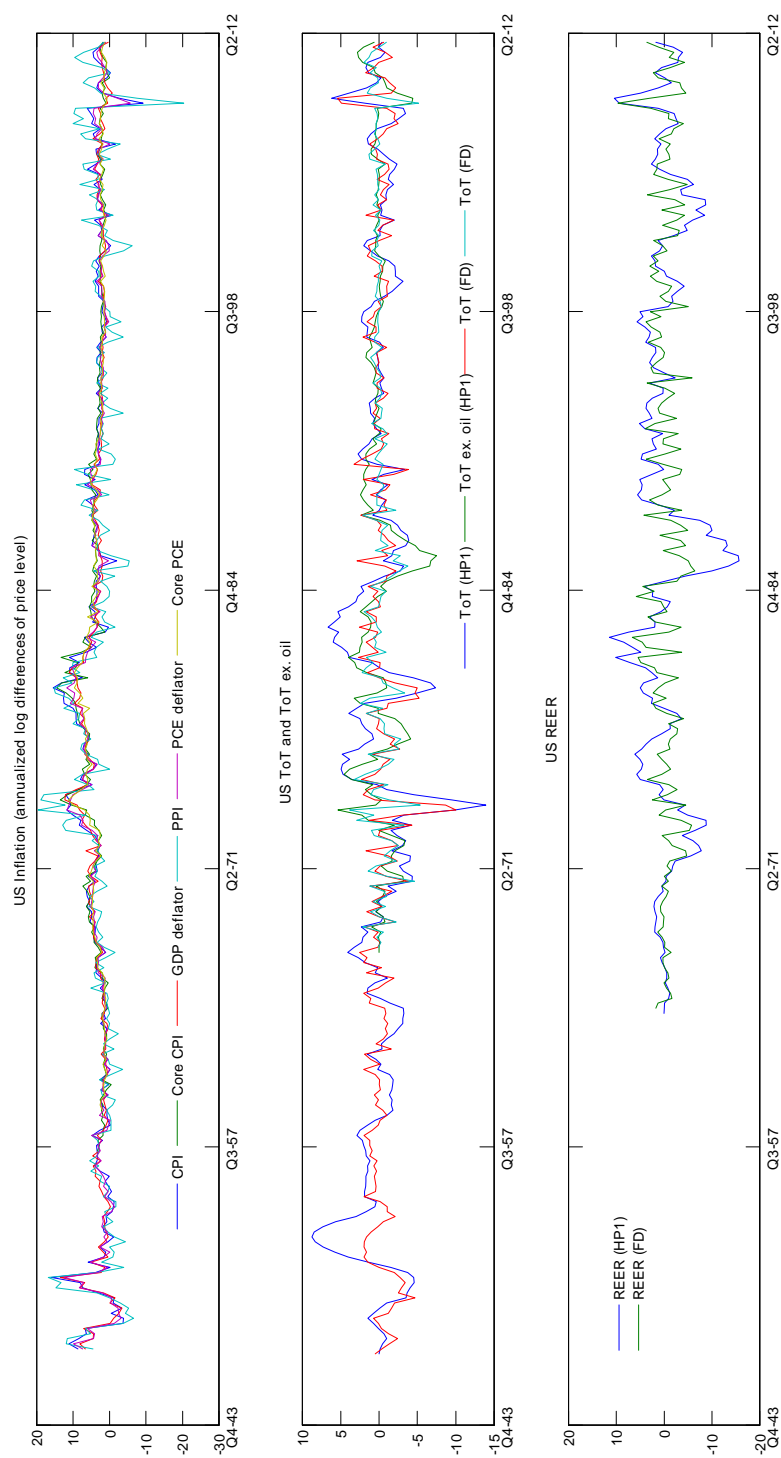


FIGURE F1. Time series plots of the data

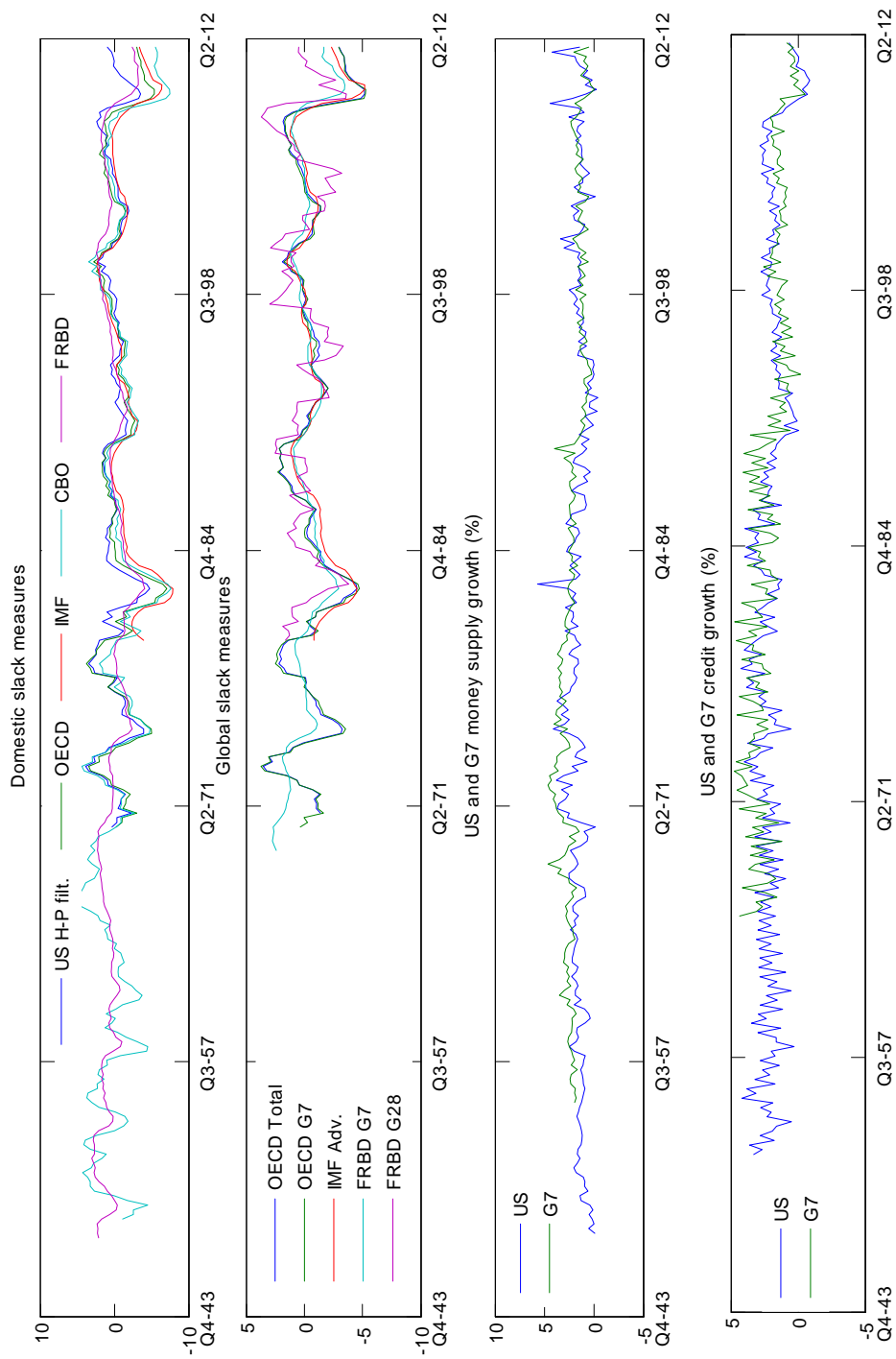


FIGURE F2. Time series plots of the data

Table F1. Relative MSFEs–1992Q1:2011Q4

Horizon	1	4	6	8	10	12	1	4	6	8	10	12
Consumer Price Index							Consumer Price Index (ex. Food & Energy)					
Autoregressive	4.704	1.881	1.334	1.164	1.065	0.958	0.295	0.418	0.626	0.828	0.954	1.036
CBO U.S. Slack	0.990*	1.042	1.059	1.070	1.100	1.168	1.135	0.974*	0.873**	0.827**	0.840**	0.870*
FRBD U.S. Slack	0.878***	0.936*	0.924*	0.936*	1.001	1.073	1.333*	0.955**	0.788**	0.737**	0.733**	0.751**
IMF U.S. Slack	1.005	1.054	1.068	1.072	1.101	1.170	1.060	1.110	1.106	1.082	1.073	1.079
OECD U.S. Slack	0.984*	1.028	1.038	1.046	1.068	1.118	1.018	0.801***	0.741***	0.739**	0.768**	0.798**
U.S. HP-filtered	0.966	0.980	1.009	1.062	1.086	1.072	0.860***	0.778***	0.811**	0.854**	0.877*	0.889*
U.S. Money Growth	0.996**	1.017	1.030	1.075	1.100	1.147	1.059	1.048	1.090	1.124	1.150	1.183
PCE Price Index							Core PCE Price Index					
Autoregressive	2.329	1.190	0.930	0.885	0.875	0.899	0.337	0.283	0.310	0.386	0.484	0.599
CBO U.S. Slack	0.999	1.026	1.041	1.051	1.069	1.082	0.949**	0.972	0.989	0.999	1.016	1.026
FRBD U.S. Slack	1.006	1.032	1.037	1.030	1.024	1.023	1.022	1.074	1.096	1.086	1.062	1.033
IMF U.S. Slack	1.005	1.024	1.032	1.038	1.055	1.067	0.997	1.016	1.035	1.045	1.048	1.043
OECD U.S. Slack	0.995	1.012	1.020	1.027	1.040	1.048	0.927***	0.920*	0.921*	0.924*	0.946	0.966
U.S. HP-filtered	0.987*	1.022	1.043	1.065	1.071	1.071	0.987*	1.007	1.024	1.041	1.046	1.056
U.S. Money Growth	1.002	1.020	1.045	1.078	1.091	1.125	0.996	1.077	1.132	1.162	1.178	1.196
GDP Deflator							Producer Price Index					
Autoregressive	0.508	0.376	0.423	0.498	0.545	0.575	19.431	7.829	4.724	3.127	2.207	1.786
CBO U.S. Slack	1.005	0.996	1.030	1.040	1.055	1.066	0.980*	1.007	1.032	1.039	1.073	1.119
FRBD U.S. Slack	0.949**	0.910**	0.962*	0.997	1.036	1.050	1.005	1.064	1.102	1.137	1.188	1.250
IMF U.S. Slack	1.037	1.057	1.059	1.048	1.057	1.065	0.996	1.024	1.053	1.082	1.143	1.227
OECD U.S. Slack	0.970**	0.951*	0.999	1.022	1.037	1.047	0.974**	0.989	1.001	0.999	1.006	1.013
U.S. HP-filtered	0.980*	0.980	1.032	1.062	1.069	1.077	0.991	1.023	1.016	1.019	0.988	0.926*
U.S. Money Growth	1.004	1.033	1.057	1.074	1.090	1.108	1.000	1.011	1.035	1.068	1.101	1.157
Sticky Price							Sticky Price ex. Shelter					
Autoregressive	0.048	0.048	0.072	0.096	0.111	0.120	0.049	0.062	0.092	0.115	0.129	0.137
CBO U.S. Slack	1.085	1.087	1.068	1.055	1.044	1.034	1.807	0.871**	0.775**	0.753**	0.748**	0.755**
FRBD U.S. Slack	1.144	0.919**	0.871**	0.788**	0.738**	0.706**	1.387	0.802**	0.735***	0.709**	0.688**	0.678**
IMF U.S. Slack	1.179	0.943	0.838	0.806	0.791	0.782	1.369	0.857***	0.759***	0.717**	0.707**	0.703
OECD U.S. Slack	1.034	1.026	1.024	1.029	1.032	1.032	1.619	0.779	0.708	0.712	0.722	0.739
U.S. HP-filtered	0.921***	0.822**	0.812**	0.841**	0.861*	0.882*	0.952**	0.808**	0.804**	0.821**	0.838**	0.855*
U.S. Money Growth	1.015	1.050	1.076	1.102	1.120	1.130	1.084	1.059	1.082	1.108	1.129	1.147

Note: This table reports the forecasting performances with an estimation sample covering 1980Q1:1991Q4 and a pseudo out-of-sample forecasting sample over 1992Q1:2011Q4. The first row of each panel shows the MSFEs of forecasts with the simple univariate AR process of inflation (restricted model) and are therefore in absolute terms. The remaining entries are the MSFEs of the forecasts under the unrestricted model relative to the MSFEs of the restricted model. Asterisks denote that the relative MSFEs are statistically different and (more accurate) than the MSFEs of the benchmark (restricted) model at 1 (**), 5 (*), and 10 (*) percent significance levels.

Table F2. Relative MSFEs–1992Q1:2011Q4

Horizon	1	4	6	8	10	12	1	4	6	8	10	12
Consumer Price Index												
Autoregressive	4.704	1.881	1.334	1.164	1.065	0.958	0.295	0.418	0.626	0.828	0.954	1.036
FRBD G7	0.977*	1.049	1.014	1.017	1.049	1.076	0.947**	0.912**	0.776**	0.740**	0.745**	0.769**
FRBD G28	0.969**	0.974	1.006	1.080	1.121	1.104	0.896**	0.819**	0.852**	0.899*	0.921*	0.927
IMF Adv.	1.005	1.050	1.054	1.046	1.041	1.059	1.018	1.054	1.058	1.039	1.023	1.014
OECD G7	0.981	1.036	1.049	1.060	1.067	1.087	0.905**	0.769***	0.680***	0.668***	0.699**	0.735**
OECD Total	0.989*	1.045	1.061	1.073	1.081	1.101	0.968**	0.816***	0.701***	0.677***	0.703**	0.738**
G7 Money Growth	1.010	0.908**	0.854**	0.832**	0.783**	0.736**	0.971**	0.864**	0.898***	0.813**	0.848**	0.860**
Core PCE Price Index												
Autoregressive	2.329	1.190	0.930	0.885	0.875	0.899	0.337	0.283	0.310	0.386	0.484	0.599
FRBD G7	0.934***	1.059	1.045	1.083	1.130	1.131	1.018	1.039	1.032	1.023	1.008	1.099
FRBD G28	0.989*	1.010	1.048	1.110	1.135	1.131	0.982*	0.995	1.041	1.093	1.103	1.094
IMF Adv.	1.006	1.013	0.993	0.977	0.966	0.960	1.005	1.015	1.016	1.020	1.010	1.001
OECD G7	0.994	1.018	1.018	1.017	1.016	1.014	0.951**	0.964	0.967	0.976	0.989	1.009
OECD Total	0.999	1.024	1.023	1.022	1.022	1.021	0.966**	0.980	0.984	0.993	1.006	1.026
G7 Money Growth	1.001	0.845***	0.852**	0.840**	0.785***	0.782***	0.991	0.932**	0.973	0.931*	0.939*	0.951
Producer Price Index												
Autoregressive	0.508	0.376	0.423	0.498	0.545	0.575	19.431	7.829	4.724	3.127	2.207	1.786
FRBD G7	1.053	0.989	0.991	1.013	1.059	1.074	1.009	1.050	1.052	1.061	1.050	1.023
FRBD G28	0.968**	0.986	1.053	1.111	1.134	1.136	0.994	1.016	1.048	1.111	1.147	1.110
IMF Adv.	0.996	0.780***	0.811**	0.887*	0.951	1.000	0.990*	1.020	1.051	1.078	1.106	1.134
OECD G7	0.992	1.001	1.026	1.026	1.029	1.029	0.964**	0.994	1.018	1.031	1.039	1.034
OECD Total	1.005	1.015	1.035	1.032	1.037	1.038	0.971**	1.003	1.027	1.040	1.047	1.041
G7 Money Growth	1.030	0.973*	0.951*	0.928*	0.915*	0.909*	1.019	0.999	1.001	1.020	1.041	1.020
Sticky Price ex. Shelter												
Autoregressive	0.048	0.048	0.072	0.096	0.111	0.120	0.049	0.062	0.092	0.115	0.129	0.137
FRBD G7	1.046	1.044	1.016	0.983	0.944	0.906	1.070	0.878**	0.822**	0.808**	0.805**	0.808**
FRBD G28	0.925***	0.872**	0.877**	0.908*	0.932*	0.952	0.959**	0.871**	0.878**	0.898*	0.918*	0.933
IMF Adv.	1.051	1.054	1.032	0.999	0.963	0.925	1.075	1.035	1.025	1.004	0.976	0.948
OECD G7	0.986*	0.973	0.997	1.014	1.017	1.011	1.614	0.731***	0.637***	0.650***	0.676**	0.706**
OECD Total	0.999	0.992	1.009	1.021	1.021	1.013	1.630	0.773***	0.664***	0.665***	0.687**	0.718**
G7 Money Growth	0.943***	0.870**	0.835**	0.866**	0.898**	0.906*	1.051	0.873**	0.857**	0.896**	0.931*	0.944

Note: This table reports the forecasting performances with an estimation sample covering 1980Q1:1991Q4 and a pseudo out-of-sample forecasting sample over 1992Q1:2011Q4. The first row of each panel shows the MSFEs of forecasts with the simple univariate AR process of inflation (restricted model) and are therefore in absolute terms. The remaining entries are the MSFEs of the forecasts under the unrestricted model relative to the MSFEs of the restricted model. Asterisks denote that the relative MSFEs are statistically different and (more accurate) than the MSFEs of the benchmark (restricted) model at 1 (**), 5 (*), and 10 (*) percent significance levels.

Table F3. Relative MSFEs—1992Q1:2011Q4

Horizon	1	4	6	8	10	12	1	4	6	8	10	12
Consumer Price Index												
Autoregressive	4.704	1.881	1.334	1.164	1.065	0.958	0.295	0.418	0.626	0.828	0.954	1.036
ToT FD	1.011	0.979*	0.989	0.996	0.997	1.011	1.031	0.988	0.983*	0.989	0.995	1.002
ToT HP1	1.009	1.009	1.041	1.047	1.030	0.993	0.945**	0.959*	0.969	0.985	0.989	0.968
ToT ex. oil FD	1.019	1.024	1.014	0.987	0.954**	0.899***	1.094	1.135	1.113	1.063	1.034	1.005
ToT ex. oil HP1	1.023	0.991	0.931*	0.874*	0.787**	0.711*	1.042	1.058	1.050	1.017	0.979	0.936
PCE Price Index												
Autoregressive	2.329	1.190	0.930	0.885	0.875	0.899	0.337	0.283	0.310	0.386	0.484	0.599
ToT FD	1.006	0.988*	0.990	0.998	0.999	1.008	0.944***	1.011	0.999	0.993	0.989	0.992
ToT HP1	1.009	1.035	1.042	1.027	0.977	0.899**	0.958**	0.960*	0.884**	0.834**	0.804***	0.756***
ToT ex. oil FD	1.025	0.998	0.962**	0.927**	0.874***	0.831***	1.057	0.958**	0.915**	0.894***	0.853***	0.844***
ToT ex. oil HP1	1.012	0.913**	0.831**	0.750**	0.645**	0.566***	0.989	0.901**	0.837**	0.792**	0.747**	0.712**
GDP Deflator												
Autoregressive	0.508	0.376	0.423	0.498	0.545	0.575	19.431	7.829	4.724	3.127	2.207	1.786
ToT FD	0.979*	0.957**	0.985*	1.005	1.007	1.005	1.020	0.996	1.013	1.019	1.037	1.035
ToT HP1	0.929***	0.932**	0.952*	0.949*	0.910**	0.852**	1.070	1.116	1.204	1.378	1.594	1.649
ToT ex. oil FD	0.985*	1.020	0.986	0.964*	0.926**	0.904***	1.081	1.060	1.126	1.152	1.233	1.303
ToT ex. oil HP1	0.976*	0.952*	0.898*	0.849*	0.782**	0.720**	1.033	1.042	1.066	1.090	1.146	1.195
Sticky Price												
Autoregressive	0.048	0.048	0.072	0.096	0.111	0.120	0.049	0.062	0.092	0.115	0.129	0.137
ToT FD	1.041	0.970**	0.978*	0.994	1.003	1.007	1.086	0.974**	0.977**	0.992	1.002	1.005
ToT HP1	0.897***	0.907**	0.966	0.993	0.991	0.973	0.910***	0.923**	0.964	0.990	0.988	0.969
ToT ex. oil FD	1.069	1.074	1.076	1.052	1.029	1.008	1.066	1.069	1.057	1.042	1.025	1.004
ToT ex. oil HP1	1.019	1.004	0.990	0.970	0.939	0.904	1.028	1.010	0.997	0.979	0.953	0.923

This table reports the forecasting performances with an estimation sample covering 1980Q1:1991Q4 and a pseudo out-of-sample forecasting sample over 1992Q1:2011Q4. The first row of each panel shows the MSFEs of forecasts with the simple univariate AR process of inflation (restricted model) and are therefore in absolute terms. The remaining entries in each panel report the relative MSFEs of the univariate forecasts with terms of trade measures. Asterisks denote that the relative MSFEs are statistically different and (more accurate) than the MSFEs of the restricted model at 1 (**), 5 (*), and 10 (*) percent significance levels.

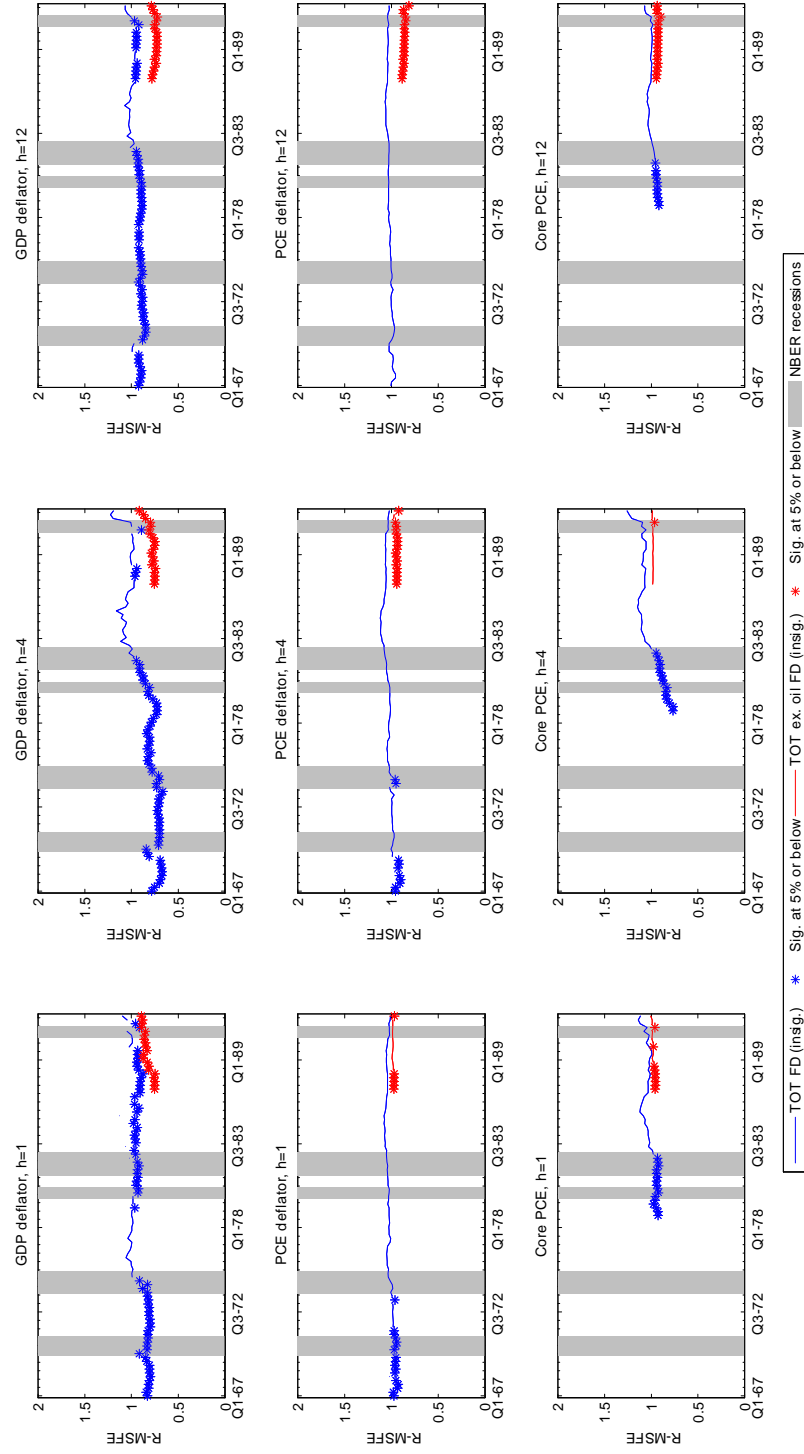


FIGURE F3. Evolution of the relative MSFEs of the forecasts with the terms of trade vs. terms of trade ex. oil (first differenced)

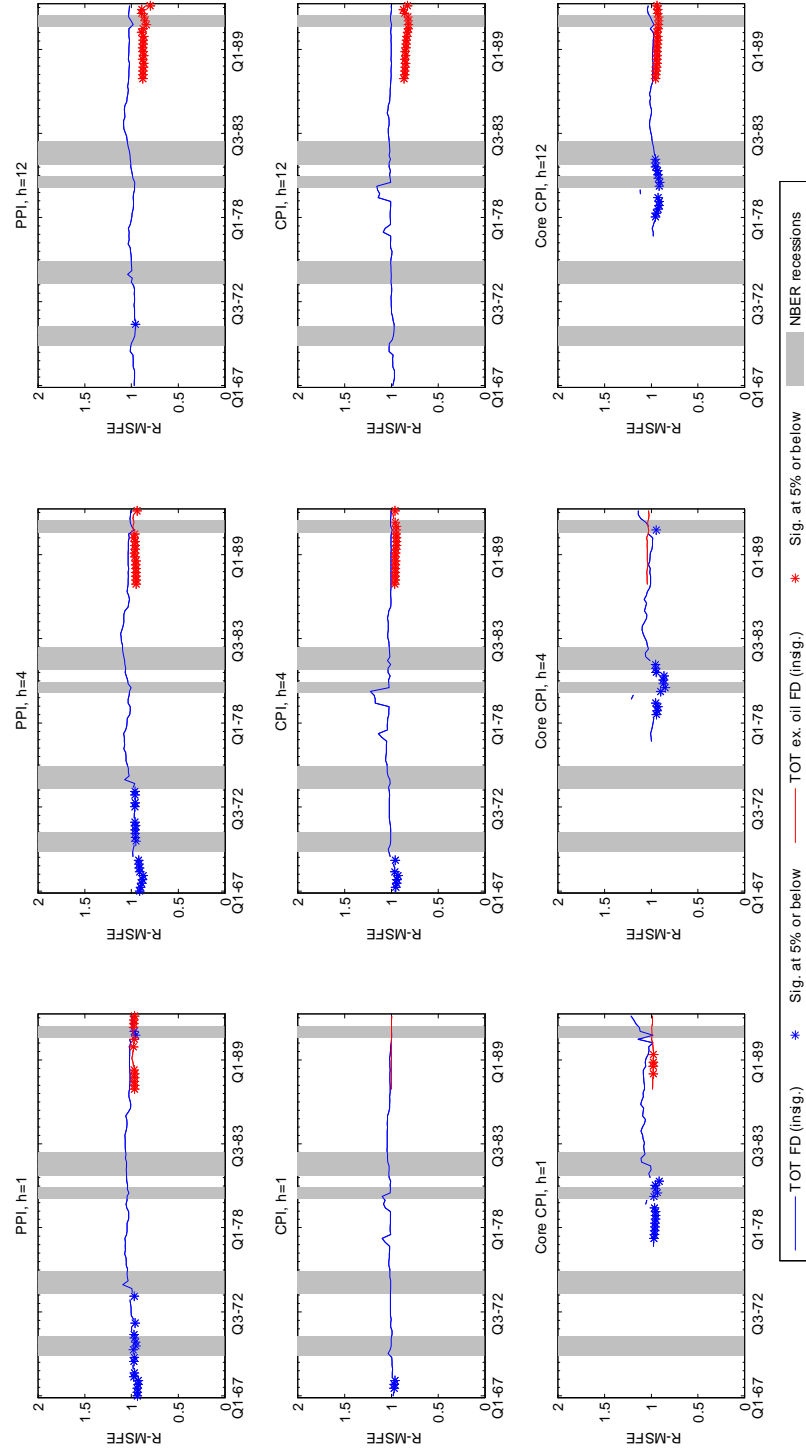


FIGURE F4. Evolution of the relative MSFEs of the forecasts with the terms of trade ex. oil (first differenced)

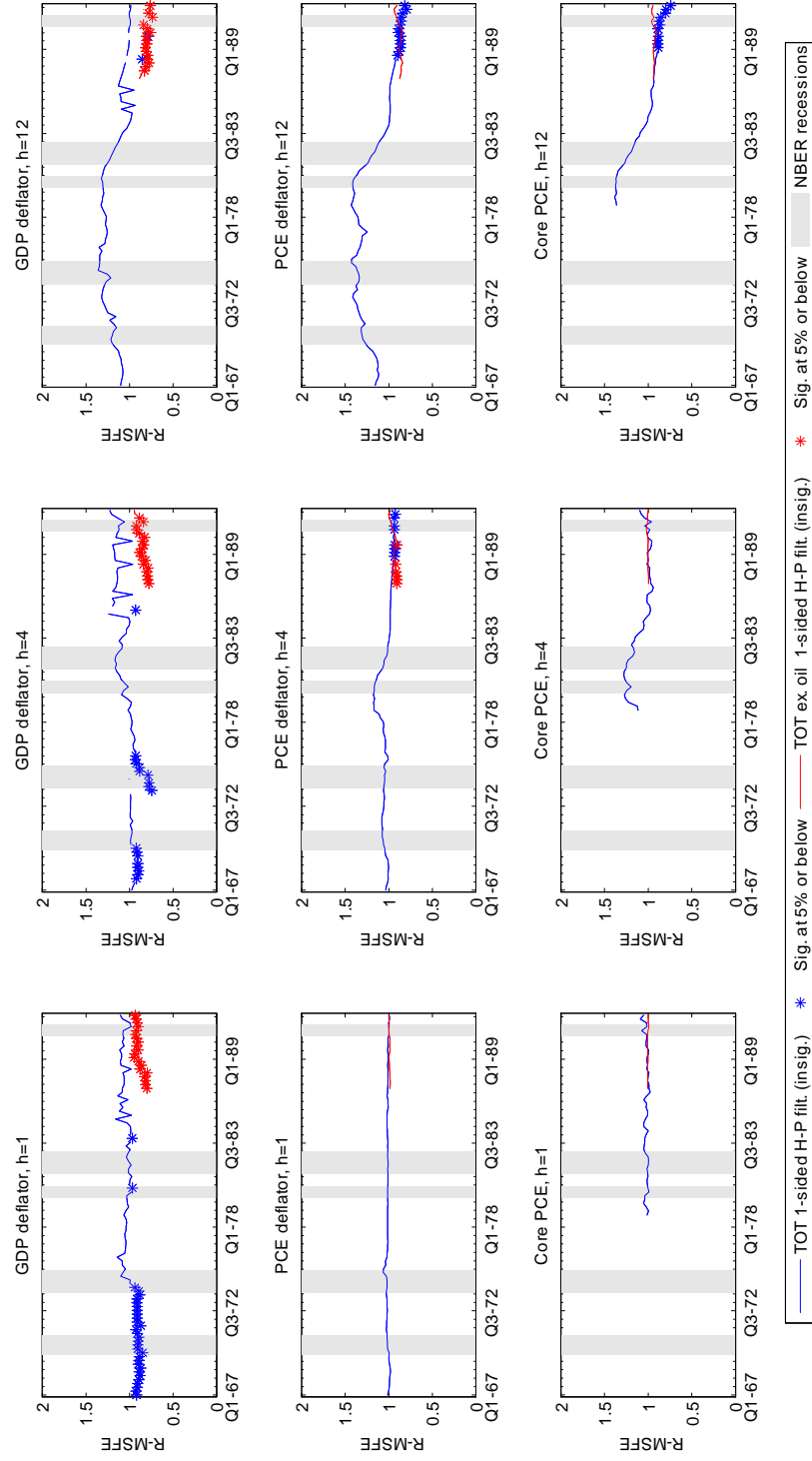


FIGURE F5. Evolution of the relative MSFEs of the forecasts with the terms of trade vs. terms of trade ex. oil (1-sided Hodrick-Prescott filtered)

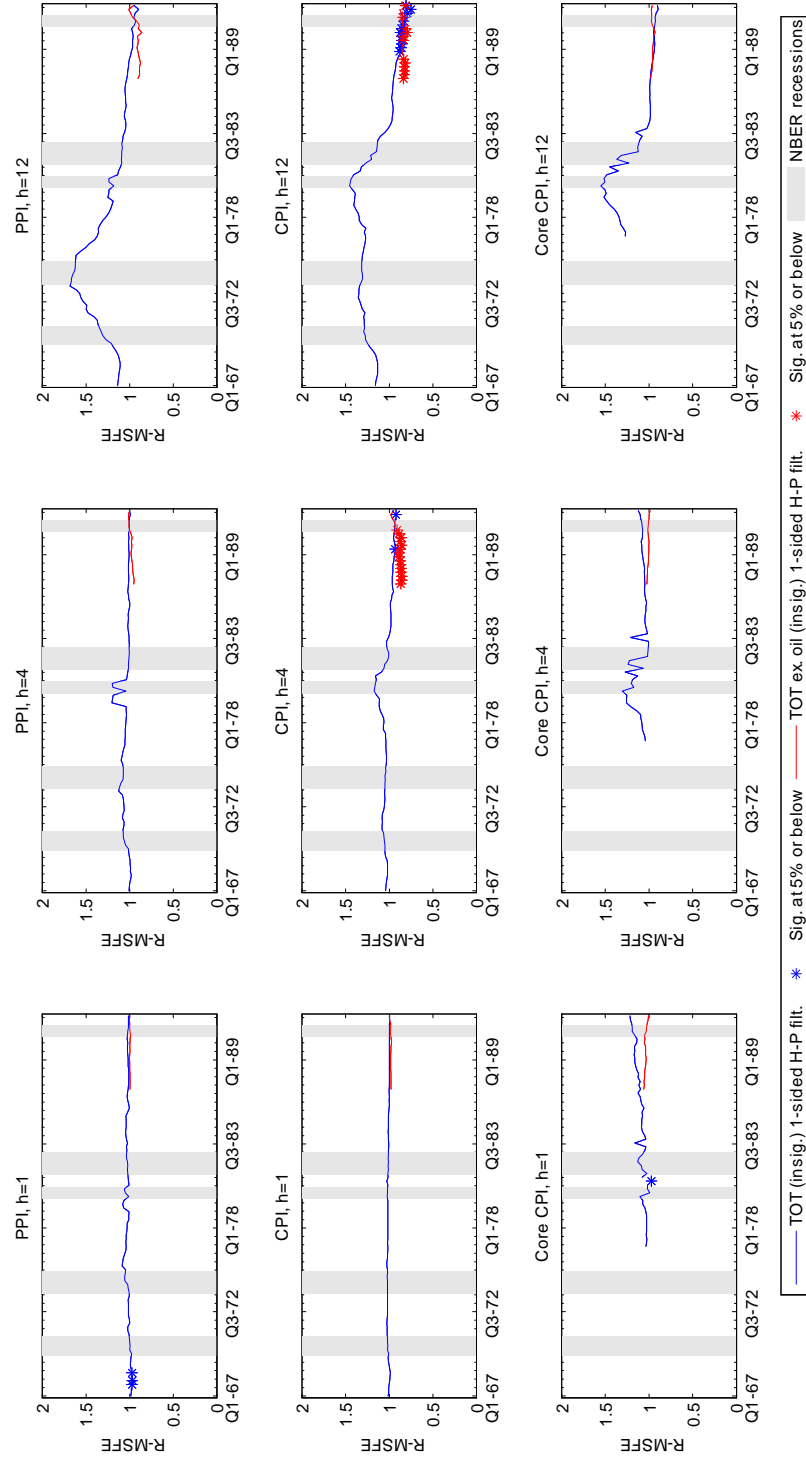


FIGURE F6. Evolution of the relative MSFEs of the forecasts with the terms of trade vs. terms of trade ex. oil (1-sided Hodrick-Prescott filtered)

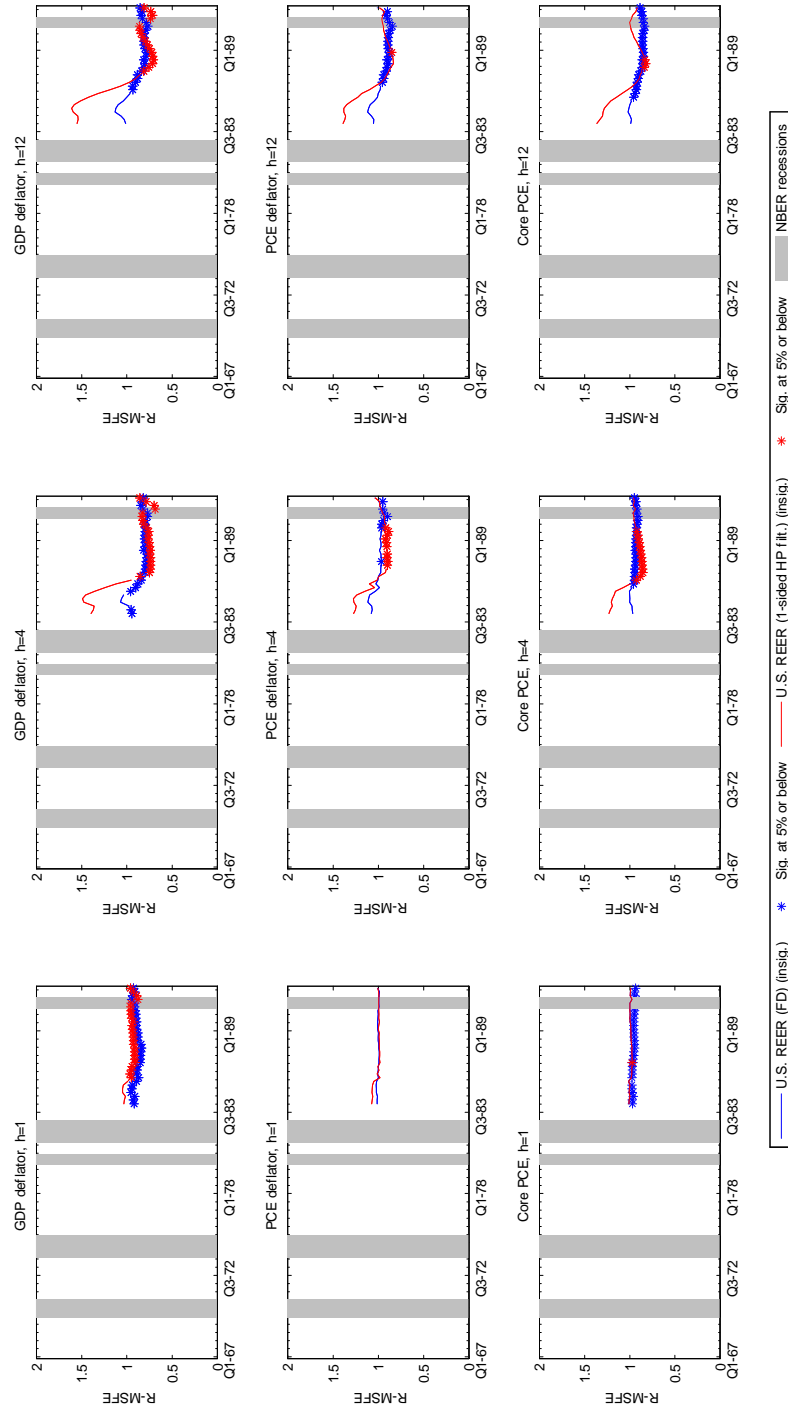


FIGURE F7. Evolution of the relative MSFEs of the forecasts with the REER (first differences) vs. REER (1-sided Hodrick-Prescott filtered)

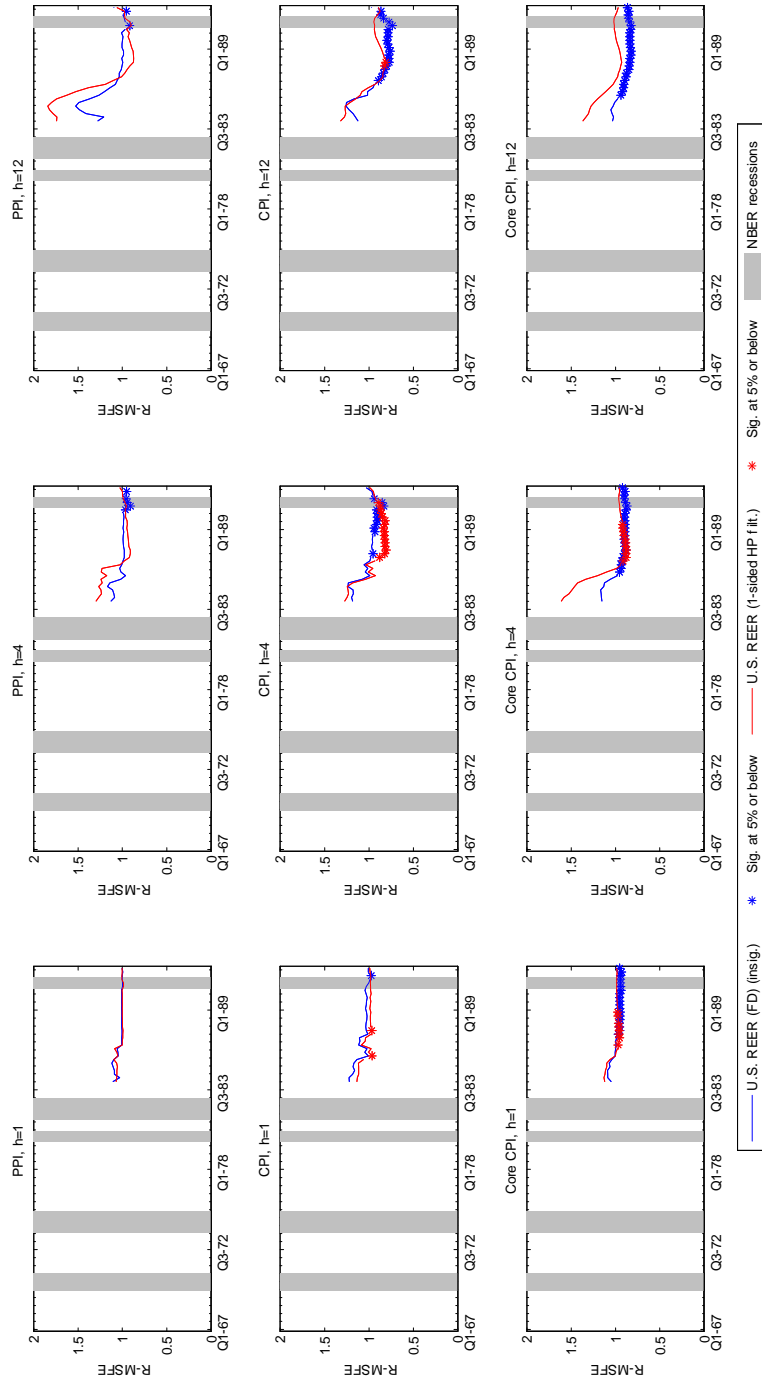


FIGURE F8. Evolution of the relative MSFEs of the forecasts with the REER (first differences) vs. REER (1-sided Hodrick-Prescott filtered)

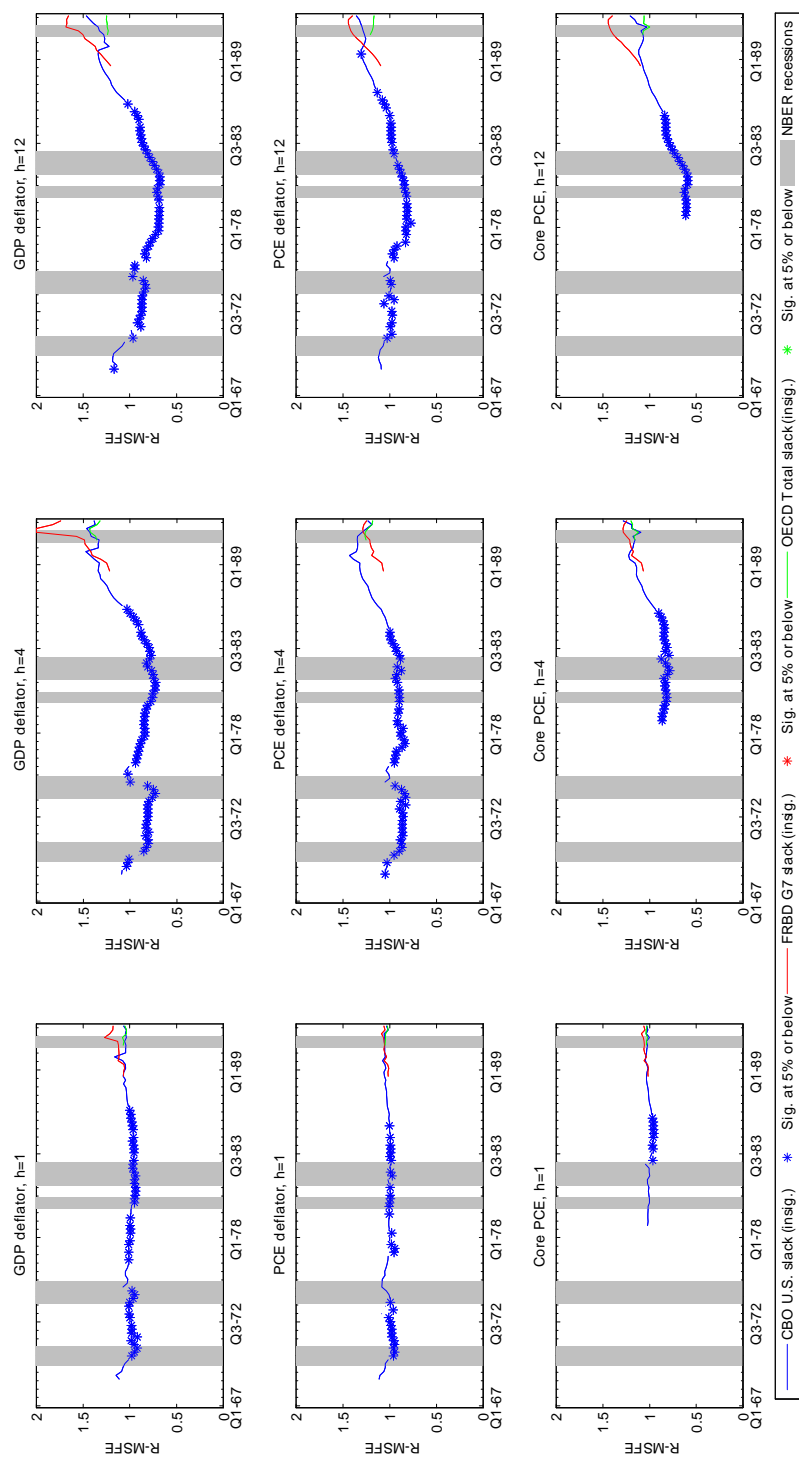


FIGURE F9. Evolution of the relative MSFEs of the forecasts with CBO U.S., FRBD G7, and OECD Total Slack

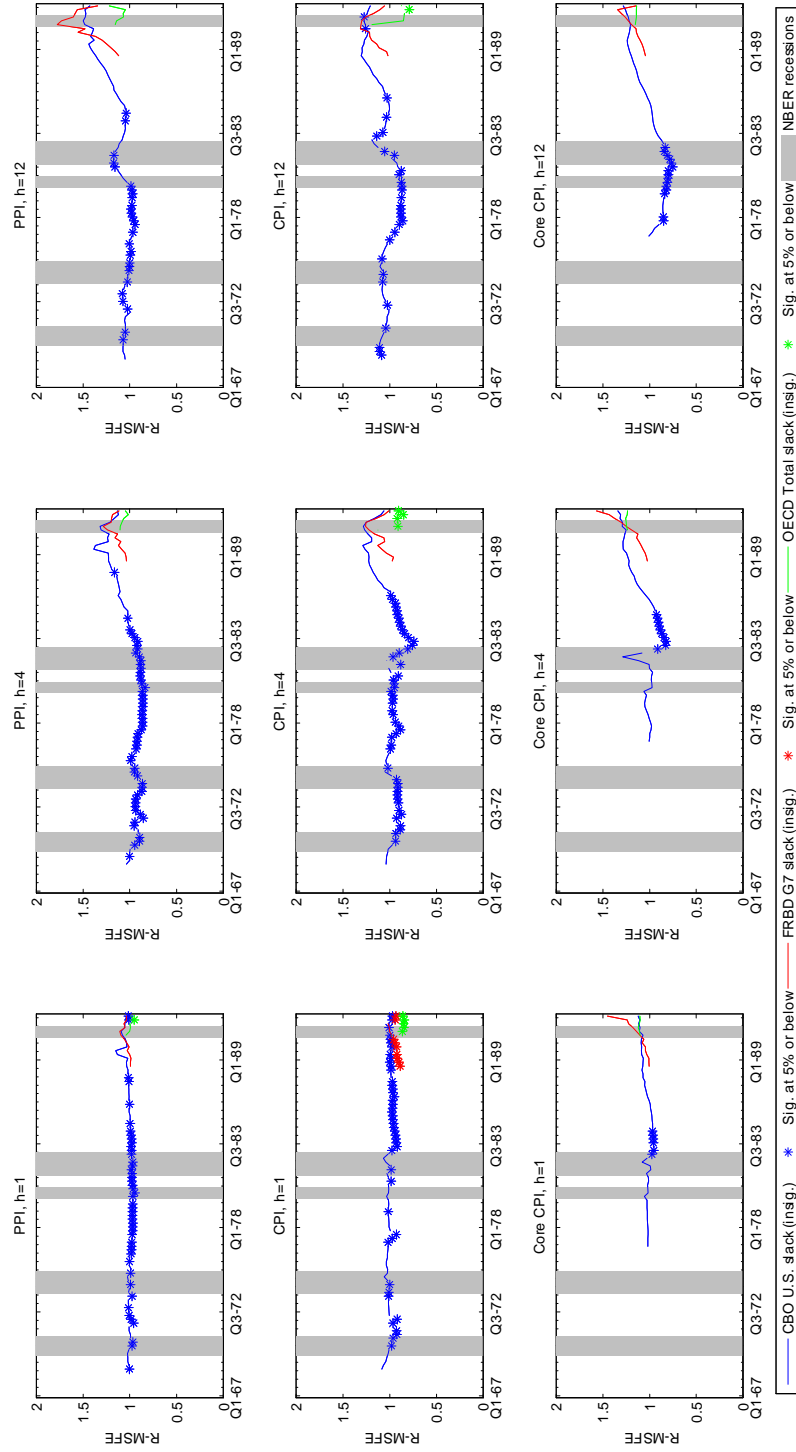


FIGURE F10. Evolution of the relative MSFEs of the forecasts with the CBO U.S., FRBD G7, and OECD Total Slack

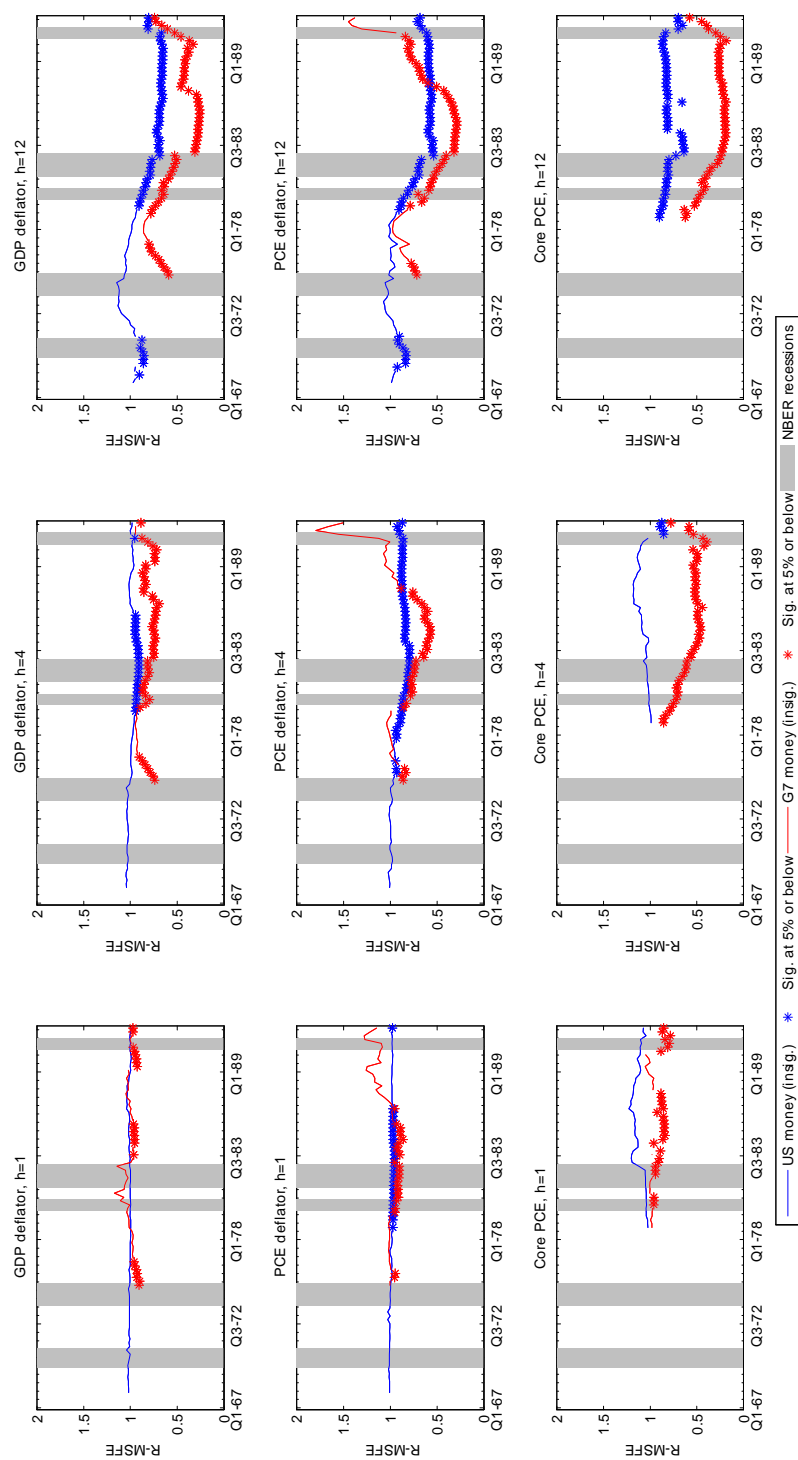


FIGURE F11. Evolution of the relative MSFEs of the forecasts with U.S. and G7 money supply growth

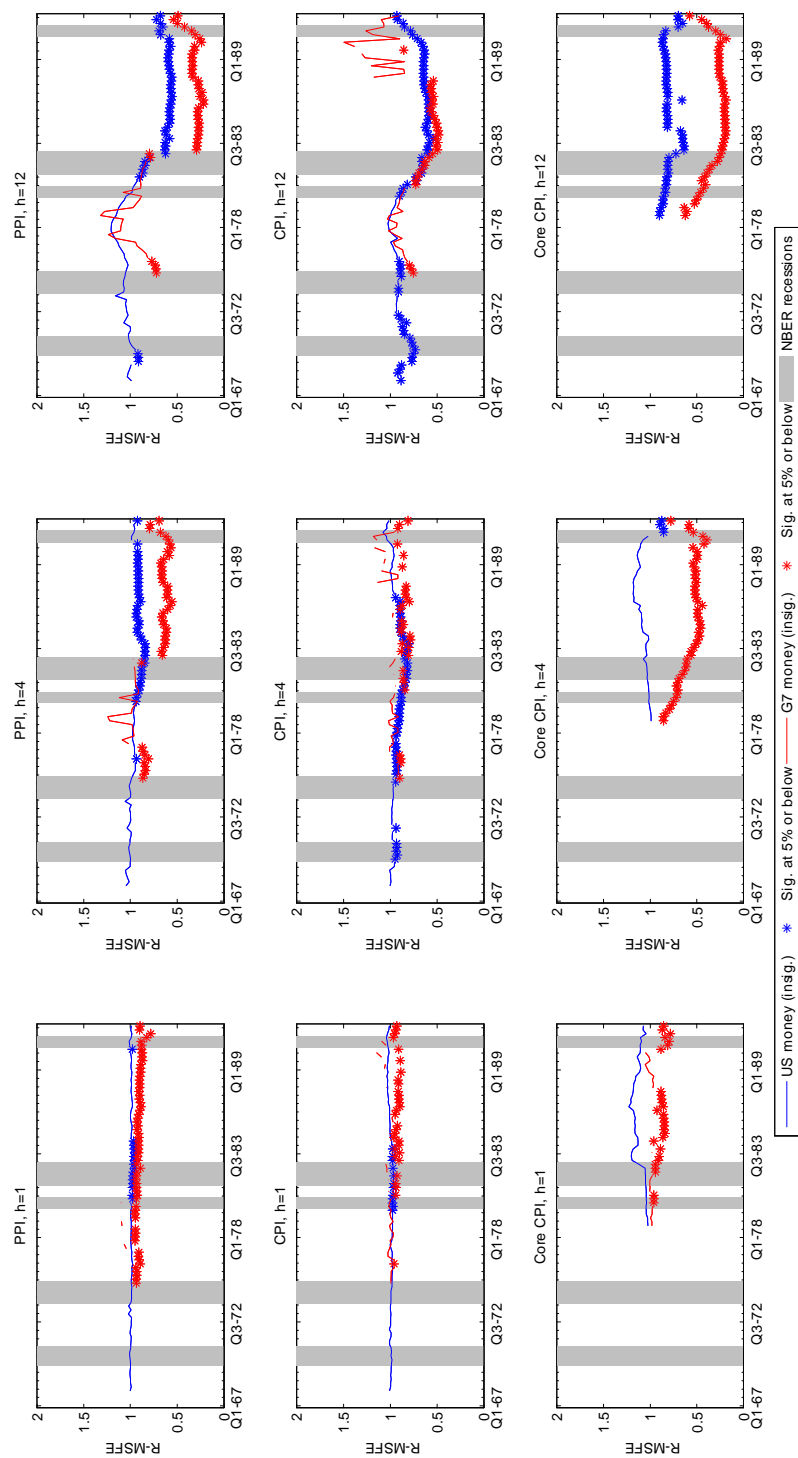


FIGURE F12. Evolution of the relative MSFEs of the forecasts with U.S. and G7 money supply growth

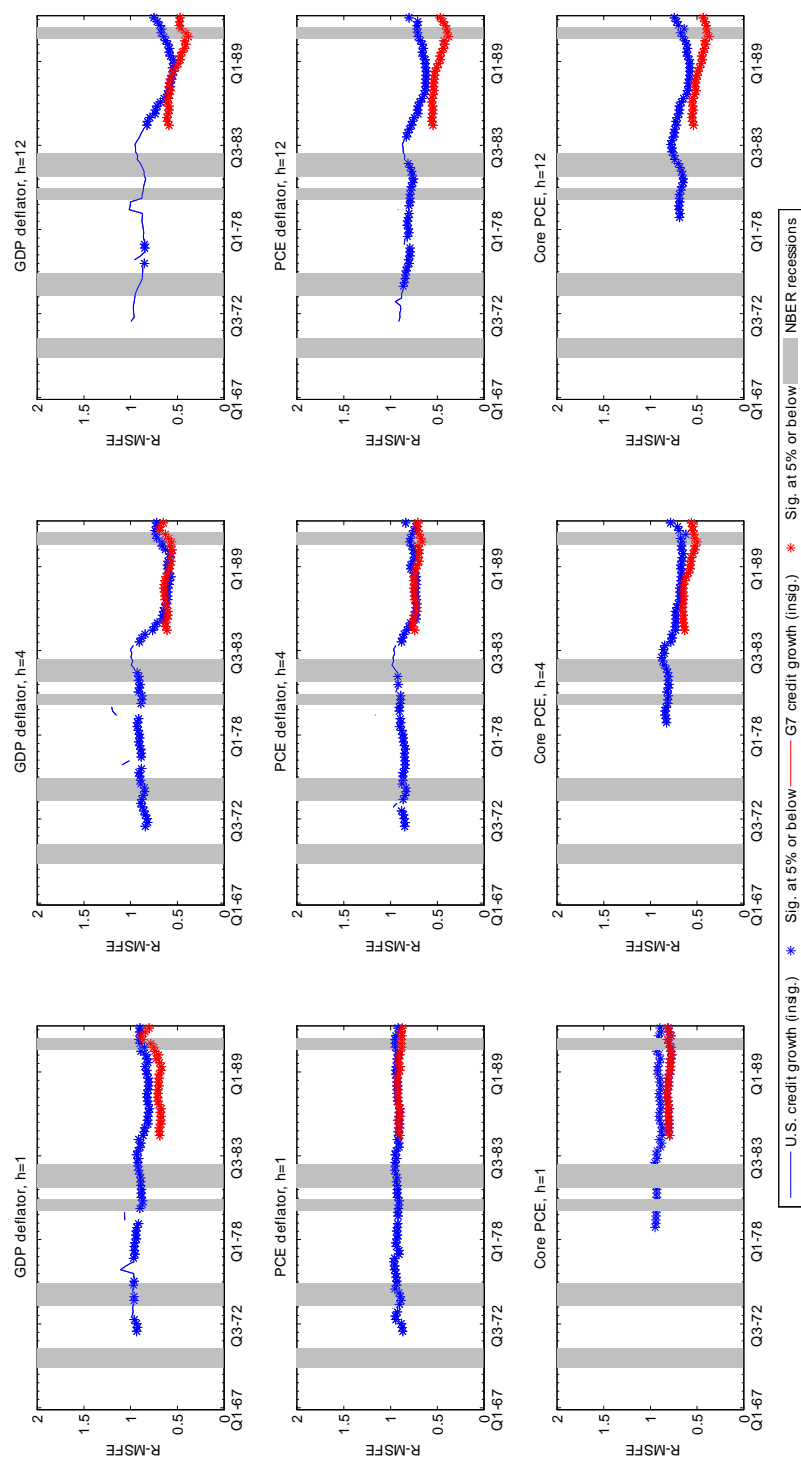


FIGURE F13. Evolution of the relative MSFEs of the forecasts with U.S. and G7 credit growth

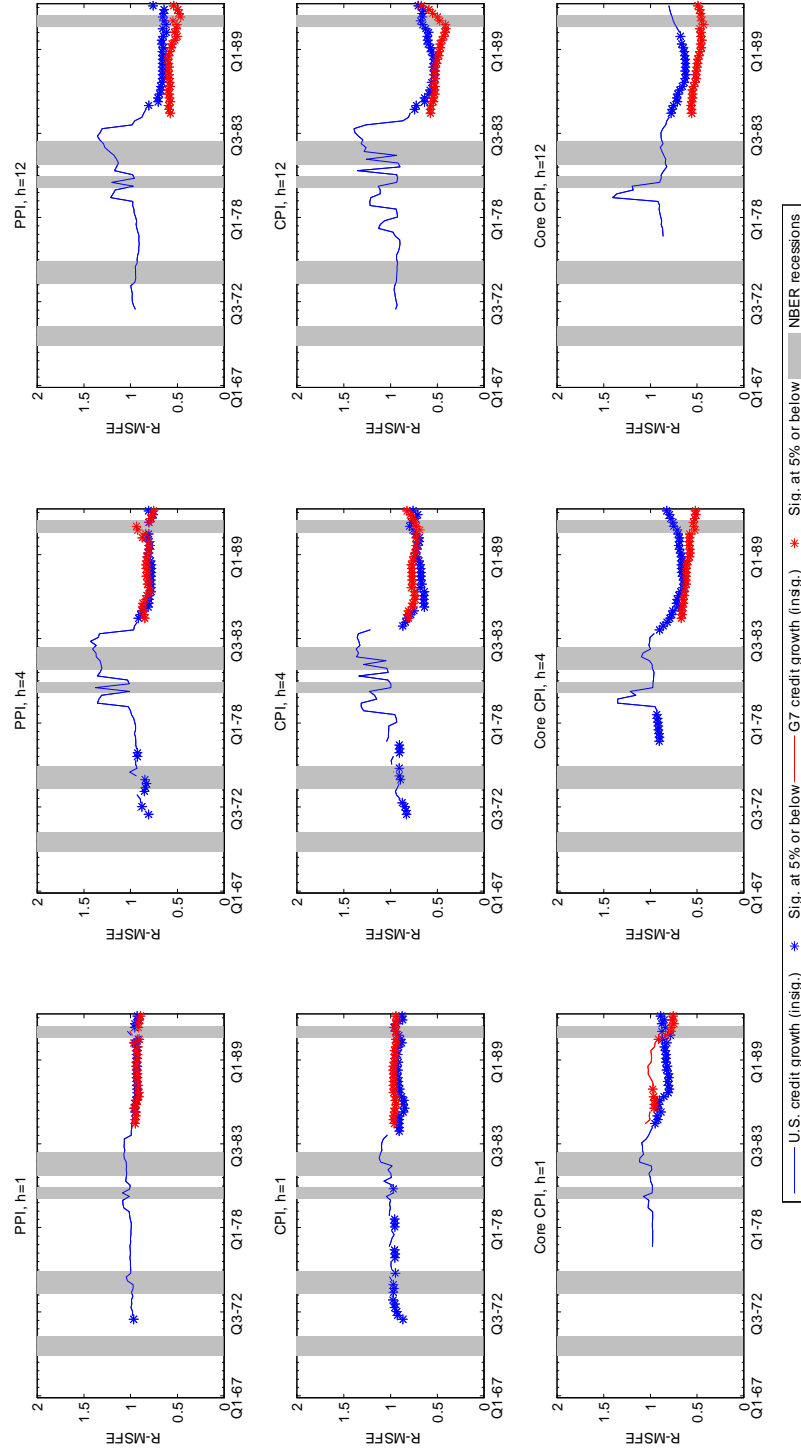


FIGURE F14. Evolution of the relative MSFEs of the forecasts with U.S. and G7 credit growth