Global Stability of In‡ation Target Policies with Adaptive Agents[¤]

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Abstract

We study a dynamic equilibrium model where agents have adaptive expectations and monetary authorities pursue an in‡ation target. We show how alternative monetary stabilization polices become more e¤ective when …scal constraints on de…cits are implemented, although they are not binding at the equilibrium target. In particular, we show that the in‡ation target equilibrium can be locally, or even globally, stable for a large class of adaptive learning schemes. We also compare alternative stabilization policies in terms of their stability properties. Commonly postulated conditional Taylor-type rules tend to be dominated by other rules, such as an unconditional Friedman-type rule.

Key words: In‡ation Targeting, Adaptive Expectations, Stability, Global Dynamics.

[¤]Some of the local stability results reported here were …rst described in Barucci, Bischi and Marimon (1998). We want to thank Emilio Barucci for his instrumental role at the start of this joint project. Previous versions of this paper were presented at Stokholm (IIES), Trieste and in the 1999 SED Meeting in Alghero. We want to thank seminar participants for their comments.

1 Introduction

As monetary policy design enters the 21st century, the more than half-century-old Friedman (1948) dictum, "rules rather than discretion," seems to de…ne the predominant view among academics and many central bankers. More speci…cally, a goal of price stability has become the norm and, to this aim two policy options dominate the debate. One is the need for …scal constraints (at least constraints on seignorage) as a way to force monetary authorities to pursue price stability. The second is the more or less explicit implementation of an in‡ation target rule. The former is seen as a commitment device while the second is seen, once commitment has been granted, as a stabilization policy. We focus on one (usually neglected) aspect of such policies: the role of expectations formation in the design of policy rules. In particular, we investigate how …scal constraints can help achieve price stability (even when there are no credibility problems) and how di¤erent in‡ation target policies can be ranked according to their stability properties in economies where private agents form their expectations adaptively.

In taking into account the role of …scal (seignorage) constraints in economies with adaptive learning agents, we follow up on the recent work of Evans, Honkapohja and Marimon (2000; E-H-M, hereafter). However, in contrast with that work, we consider alternative policies for the central bank (they only consider …xed seignorage …nancing) and a wider class of (deterministic) learning rules for private agents. In particular, our analysis of alternative stabilization policy rules aims at sheding some light on the discussion of how in‡ation target policies should be designed. Our analysis of a wide class of learning rules aims at taking into account for the fact that, when observed in‡ation di¤ers from the …xed (trivially stationary) target, private agents are likely to place more weight on recent data. Taking this broader perspective allows us to study how di¤erent parameters a¤ect the price stability under alternative rules. For example, we show how …scal constraints may enhance price stabilization in ways that could not be captured either by rational expectations models or by adaptive learning models with decreasing gain (such as least squares learning, as studied in E-H-M).

We show how di¤erent monetary instruments are equivalent to the use of a single intermediate instrument determining the ex-post real return on money. In setting the value of such an instrument (e.g., what would correspond to setting the current interbank rate) the central bank may condition on current information (i.e., deviations from an output target), but has also to forecast the demand for money which, in our model, reduces to forecast "private agents' expectations." Di¤erent in‡ation target policies di¤er on how the government conditions on past data and on its beliefs regarding private agents' expectations. The policy that we identify as the "optimal" policy is the one that uses all available information and, therefore, conditions on observed deviations. Such a policy is consistent with rational expectations, in the sense that the monetary authority (assumed to be fully committed to its policy) forecasts that private agents' expect that the target is achieved in the short run. Such a policy is of the form of the in‡ation target policies proposed by Svensson (1997) and others. However, under our policy the target is only one of many possible rational expectations equilibria. In fact, as Benhabib et al. (1999) have recently shown, Taylor rule policies may result in indeterminacy and, in particular, in paths that diverge from the target (when policy is "active," see Section 2). Along these paths, as it often happens with observed series, in‡ation is autocorrelated and deviations from target can not be accounted for as simple stochastic innovations.

What should in‡ation target policy be when deviations from target are not innovations? A …rst possibility is to think that the "optimal"policy remains in place. Implicitly, this is the view adopted by the existing literature on Taylor rules (see, for example, McCallum 1997, Mishkin and Posen 1997, and Clarida, Galí and Gertler 1997a). A second possibility is to go back to Friedman's recommendation and postulate an unconditional policy consistent with the long-run objectives. Finally the central bank can try to "forecast how private agents forecast." This, of course, is not a closed—or well de…ned— possibility and it raises a number of interesting issues. We …nd that, if the central bank succeeds at such forecasting game, then, as in the rational expectations case, the target should prevail in the short run and the best forecast of private agents' beliefs is the same target (see Section 2). However, central banks may not be that farsighted, they may simply postulate a certain amount of inertia on how private agents forecast. As a canonical example we postulate a simple (…xed) adaptive rule as a conditional in‡ation target rule. Studying and comparing the performance of the three rules, in an economy where agents' expectations are adaptive, is the central theme of this paper. For all three rules there is, of course, a misspeci…cation problem: The central bank does not implement a rule that is fully consistent with how private agents learn, nor do private agents postulate learning rules fully consistent with the actual law of motion implied by the central bank policy.¹ Nevertheless, we show that for a wide range of parameters the in‡ation target is a stable equilibrium of the corresponding adaptive process.

We …nd that, when policy is "active," under learning the in‡ation target is more stable when the stationary rational expectations equilibrium is a (locally unique) determinate equi-

¹See Sargent (1999) for a discussion of adaptive models with misspeci...ed beliefs.

librium. In this respect our work reinforces and complements the (contemporaneous) work of Bullard and Mitra (1999), who also study the E-stability of in‡ation target policies.² We study a somewhat narrower set of policies than they do, and we provide a full characterization of stability results, not only by considering local stability of a wide class of constant gain rules, but also by considering associated global stability properties.

It is in the global analysis where this paper breaks more novel ground. First, by showing how …scal constraints may a¤ect the global stability of the target, and second, by making use of some new results on global bifurcations. 3

Our exercise provides a better understanding of how three basic parameters interact with and a¤ect price stability. Two are, to a large extent, policy parameters: (i) how low the in‡ation target is set in relation to the in‡ation level at which there is no demand for money, and (ii) the tightness of the …scal constraint. The remaining parameter is endogenous to agents' learning process: (iii) how much weight they place on previous period observed information (i.e., the size of the gain or tracking parameter). In addition, we show how the three (seemingly similar) policies can result in quite di¤erent dynamics. As a result, we can provide (local and global) stability rankings. We show that, in these stability rankings, what appeared to be the "optimal" policy on other grounds actually tends to be dominated by the alternative policies. In particular, Friedman's unconditional rule performs remarkably well as stabilization policy. This may provide a rationale for the observed fact (see Clarida et al. 1997b) that central banks appear to react much less aggressively to incoming information than standard analyses of Taylor rules suggest.

The paper is divided into two important sections. Section 2 develops the model while Section 3, the bulk of the paper, contains the local and global stability results.

2 In‡ation target policies

In this section we …rst consider a general monetary model of in‡ation targeting. In the next subsection we provide a speci…c cash-in-advance interpretation of the model.

The consolidated intertemporal government budget constraint takes the form

$$
M_{t+1}^{s} + B_{t+1}^{s} = p_t g_t i p_{t\lambda t} + M_t^{s} + B_t^{s} l_t
$$
 (1)

 2 For a detailed account of E-stability theory see, for example, Evans and Honkapohja (2000)

³See, for example, Mira et al.1996, Abraham et al. 1997, Bischi et al. 1998 for an introduction to the these results on contact bifurcations.

where g_t is government expenditures, $\mathsf{p}_{\mathsf{t}\dot{\mathsf{t}}}$ is tax revenues, $\mathsf{M}_{\mathsf{t}+1}^{\mathsf{s}}$ and $\mathsf{B}_{\mathsf{t}+1}^{\mathsf{s}}$ are the supplies of money and government bonds, respectively, at the end of period t_i and I_t is the nominal rate of return on bonds (contracted in period t_i 1 at that rate). It is assumed that the sequence of intertemporal budget constraints satis…es a transversality condition, and, therefore, that the government satis…es its present value budget constraint. It is convenient to express (1) as

$$
M^s_{t+1\;\, \textrm{i}}\ \ M^s_t=p_t d_t
$$

where

$$
d_{t} = g_{t i} \dot{z}_{t} + \frac{B_{t}^{s}}{p_{t}} I_{t i} \frac{B_{t+1}^{s}}{p_{t}} \dot{g}_{t i} \dot{z}_{t} + b_{t}^{s} R_{t i}^{b} \dot{b}_{t+1}^{s}
$$
(2)

In the last equality debts and rates of return are speci...ed in real terms. In particular, R_t^b is the realized real rate of return on bonds. With this compact formulation, d_t can be identi...ed as the instrument used to implement the target (although in practice, changes on the right hand side of (2) correspond to open market operations, interbank rate interventions, etc.) While it may be important for policy design, in our model the exact form through which d_t changes is not relevant for the dynamic exects of the policy.⁴

The money market equilibrium is simply given by $\mathsf{M}^\mathsf{d}_{\mathsf{t}+1} = \mathsf{M}^\mathsf{s}_{\mathsf{t}+1}.$ Denoting real balances by $m^d_{t+1}=\frac{M^d_{t+1}}{p_t}$ and gross in‡ation by $V_{t+1}=\frac{p_{t+1}}{p_t}$ the intertemporal equilibrium condition reduces to

$$
m_{t+1}^d = \frac{m_t^d}{\mu_t} + d_t
$$
 (3)

We consider economies where the demand for real balances takes the form

$$
m^d_{t+1}=m^d(\mathtt{M}^e_{t+1})
$$

where $\mathbb{W}_{\mathfrak{t}+1}^{\rm e}$ is the agents' expected in‡ation

2.1 Introducing in‡ation target policies

An in‡ation target policy speci...es a desired level of in‡ation together with a level of d_t as a function of the available information in period t: We consider recursive policies. More speci…cally, consistent with the intertemporal equilibrium map (3), we consider policies of the form $d_t = d^{(P)}(m_t^d)$. Furthermore, if demand functions are known, these policies take

⁴ Implicitly we assume that within equivalent policies resulting in the same d policy there is (local) Ricardian equivalence; that is, present value considerations do not discriminate among these equivalent policies.

the form $d_t = d^P(\mathcal{H}_t^e)$. It follows that realized in‡ation is given by

$$
V_{4t} = A^{(P)}(V_{t}^{e}; V_{t+1}^{e}) \quad \frac{m_{\psi}^{d}(V_{t}^{e})}{m^{d}} \quad (4)
$$

Notice that, with the assumption that private agents have rational expectations, equation (4) reduces to $\mathcal{U}_t = A^{(P)}(\mathcal{U}_t; \mathcal{U}_{t+1})$: That is, we can derive an equilibrium map, $\tilde{A}^{(P)}$; such that rational expectations equilibrium paths are those satisfying

$$
V_{4t} = \tilde{A}^{(P)}(V_{4t+1})
$$

Using equation (3), in‡ation target policies take the form

$$
d^{(P)}(m_t^d) = E_t^g m_{t+1}^d \; j \; m_t^d = M^d \tag{6}
$$

where $\mathrm{E_{t}^{g}m_{t+1}^{d}}$ denotes the (government) expected demand for real balances conditional on the available information at the beginning of period t. That is, the resulting policy is conditional on past and expected future real balances.

To see the sense in which these policies are of the type of those proposed by Taylor (1993) and Svensson (1997), and estimated by Clarida, Galí and Gertler (1997a, 1997b), let $R^{\alpha} = 1 = M^{\alpha}$, $m^{\alpha} = m^{d} (M^{\alpha})$; and $d^{\alpha} = \frac{1}{M^{\alpha} + 1} M^{\alpha}$. Then equation (6) takes the form

$$
d^{(P)}(m_t^d) = d^{\pi} + {}^{\dot{I}}E_t^g m_{t+1 \, \dot{I}}^d \ m^{\alpha}^{\dot{\alpha}} + R^{\pi \, \dot{I}}m^{\alpha} \, {}^{\dot{I}} \ m_t^d^{\dot{\alpha}}: \eqno(7)
$$

That is, the central bank's optimal reaction is to increase the money supply if either the expected demand for real balances is above the target or the realized one is below the target, so as to adapt to any expected deviation from target or adjust for any experienced deviation from target. More speci...cally, in the special (linear) case ${\sf m}^{\sf d}(\mathcal{W}^{\sf e}_{t+1})= {\sf b}_{i}$ $\mathcal{W}^{\sf e}_{t+1}$; equation (7) can be written as

$$
d^{(P)}(\mathcal{H}_t^e) = d^{\alpha} + \frac{f}{4} \mathcal{H}^{\alpha} i \mathbf{E}_t^g \mathcal{H}_{t+1}^e + R^{\alpha} \left[\mathcal{H}_t^e i \mathcal{H}^{\alpha} \right];
$$

showing that the government reaction should be to increase the money supply above the target level if either it expects private sector's forecasted in‡ation to be below the target or if past expectations of in‡ation were too high. Notice that, as long as higher expected in‡ation results in lower output, a positive deviation $[\mathcal{W}^e_t \mid \mathcal{W}^u]$ corresponds to a realized value of output below the target. In other words, under $d^{(P)}$ rules, monetary authorities adapt to forecasted money demands and to realized output gaps.

However, as it can be seen from equation (6), with such a feedback rule the rate of return on money $(R_t = 1 = 4_t)$ satis…es R_t _i $R^{\pi} = {^1}m_{t+1}^d$ _i $E_t^gm_{t+1}^{\pi} = m_t$: In other words,

realized in‡ation di¤ers from target in‡ation only if the government miscalculates the private sector's demands. In fact, when the government knows the money demand function, the target is achieved—immediately—as long as the government accurately forecasts the private sector's expectations of in‡ation. This also means that the forecast consistent with rational expectations is $\mathrm{E}^{\mathfrak{g}}_{\mathfrak{t}}\mathfrak{u}_{\mathfrak{t}+1}=\mathfrak{A}^{\mathfrak{s}}$, which results in the "optimal" target policy

$$
d_t^O=d^O(m_t^d)~\hat{~}~m^d(\mathcal{V}^{\alpha})~i~R^{\alpha}m_t^d=d^{\alpha}+R^{\alpha}~i~m^{\alpha}~i~m_t^{d^{\alpha}};
$$

where the money supply is constant except for deviations of realized real balances from their target level (or output deviations, in the constant velocity case). Furthermore, consistency with rational expectations also implies that $E_{t_{i-1}}^{-1}d^O(m_t^d)^\top = d^*$. In other words, the expected money growth must be the constant growth implied by the desired in‡ation target. The constant growth of money rule $dⁿ$ is, in fact, the rule proposed by Milton Friedman, who explicitly advocated "rules rather than discretion" and also advocated designing short-run rules in terms of long-term objectives and not in terms of discretionary reactions to economic ‡uctuations (e.g., Friedman 1948). For this reason we shall refer to the constant policy d¤ as the Friedman policy d^F , given by

$$
d^F_t = d^F \ \dot{\ } \ m^d(\textbf{W}^n) \ i \ \ R^m m^d(\textbf{W}^n) = d^n \ddot{\ }
$$

Such a policy is not optimal in the sense that it does not make use of all available information as the conditional policy d $^{\rm O}$ (m $^{\rm d}_\text{t}$) does. But, as we have seen, the conditional policy should only react to unexpected deviations of ${\sf m}_{\sf t}^{\sf d}$. In particular, if the government has been following such policy and private agents have rational expectations, then it should be the case that: ${\sf m}^{\sf d}({\tt W^e_t})={\sf m}^{\sf d}({\tt W^e})={\sf m}^{\sf n}$ and, if there are no other sources of uncertainty, this implies that $d^O(m_t^d) = d^{\alpha}$:

Indeterminacy, policy activism and consistency with rational expectations

Under both policies, O and F; there is, in general, a continuum of rational expectations equilibria (REE) and two stationary rational expectations equilibria (SREE); i.e., two …xed points of $\tilde{A}^{(P)}$. In particular, under the O policy the two SREE are \mathbb{V}^{α} and $\mathsf{b} = (1 + \mathbb{V}^{\alpha})$, while under the F policy the two SREE are \mathcal{U}^{α} and $b=\mathcal{U}^{\alpha}$. Notice that F corresponds to the standard hyperin‡ation model of a constant de…cit …nanced through seignorage, and the two SREE re‡ect the existence of two in‡ation-tax levels raising the same revenues (i.e., a version of the La¤er curve). Furthermore, μ^* should be the lower steady state in‡ation rate, otherwise the target policy cannot be optimal. In fact, these models have a La¤er curve, and the two SREE generate the same revenues, but higher in‡ation is associated with lower

savings and lower welfare. For the policy F this requires $b > \frac{1}{4}a^2$. Similarly, $\frac{1}{4}a^2$ is the lower SREE in‡ation under the policy O if and only if $b > \frac{u}{q}(1 + \frac{u}{q})$, a more stringent condition than under F.

It is convenient to consider the inverse map of equation (5), say ' \tilde{A}^{i} . In fact, provided that ' ^{(P)0}(¼) > 0; if π is a SREE and ' ^{(P)0}(π) > 1 then the corresponding target policy is called active and the corresponding SREE is determinate, while if ' $^{(P)0}(\pi) < 1$ then the policy is called passive and there is indeterminacy, in the sense that a continuum of REE have a long-run in‡ation of $\frac{\pi}{l}$; i.e., a continuum of solutions of (5) with $\frac{\pi}{l}$! $\frac{\pi}{l}$ (see e.g. Leeper 1991 or Benhabib et al. 1999). It is easy to see that under any of the two policies we have ' $^{(P)0}(\%)>0$ and, provided that \mathcal{U}^{α} is the lower in‡ation SREE, ' $^{(P)0}(\mathcal{U}^{\alpha})>1.$ The high SREE is, in contrast, indeterminate and, correspondingly, the O policy is passive at $\frac{b}{1+\mathcal{H}^\text{H}}$ while the F policy is passive at $\frac{b}{\mathcal{H}^\text{H}}$: However, at high in‡ation SREE, as well as along the REE hyperin‡ationary paths approaching them, the government should realize that its target policy is not being achieved and, therefore, the rationality of the policy should be questioned. In other words, these paths are not fully consistent with rational expectations on the part of the government.

What should the government do if it observes m_t^d 6 m^π ?⁵ In the following we explore several plausible option, but we do not provide a complete answer to this question. We …rst consider the case where the government simply follows the "optimal" policy O even when output (i.e., real balance) deviations are autocorrelated. However, Friedman's implicit criticism of conditional policies as possibly being too "over-reactive" may apply to this case and, therefore, we also consider the unconditional policy F.

Policies based on forecasts of private agents' forecasts

Facing deviations from rational expectations, the government may want to infer how private agents forecast in‡ation. As we have said, if the government succeeds at "learning how private agents learn," then the resulting in‡ation must be the target, but then private agents' forecasts (forecasting rules) may be a¤ected by the corresponding shift to the announced target. This problem is similar to that of using "good predictors" of in‡ation as a guide for monetary policy as some people have proposed (see, for example, Barsky 1993). As Woodford (1994) has argued such "nonstandard indicators" su¤er from the Lucas critique problem: As much as they are "good predictors," if they are used in the design of policy

⁵ In a stochastic model, the question is what should the government do when, at some con...dence level, it infers that the predictions of private agent are not consistent with rational expectations, given the government policy.

then they should cease to be good indicators.

Let us assume that government's ability to accurately predict how private agents forecast is limited. In particular, since a broad class of learning rules show some degree of inertia,⁶ a benchmark option to consider is that the government postulates that inertia persists; i.e., $E_t^g m_{t+1}^d = m_t^d$:

Inertia in private agents' forecasts results in autocorrelated deviations from target. In particular, notice that if agents update their estimates of in‡ation according to an adaptive rule of the form

$$
M_{t+1}^e = M_t^e + \Phi_t (M_{t+1} + M_t^e) \tag{8}
$$

with \mathcal{P}_t 2 (0; 1); \mathcal{P}_t ¼ 0 (or \mathcal{P}_t & 0 as it is the case when they use standard OLS techniques), then the government is almost right (in the limit) in postulating that inertia persists, although they could choose better predictors of private agents' forecasts—namely, the same rule (8)! Postulating that (one-period) inertia persists, we get an in‡ation target policy of the form

$$
d^I(m_t^d) \wedge m_t^d{}_i\ R^m m_t^d = \frac{\mu_{\frac{M^{\alpha}}{4}i}\ 1}{\frac{M^{\alpha}}{4}}m_t^d = d^{\alpha} + (R^{\alpha}{}_i\ 1)^{\dot{I}}m^{\alpha}{}_i\ m_t^d^{\dot{\mathbb{C}}}.
$$

For $\mathcal{U}^{\alpha} > 1$ (i.e., $\mathbb{R}^{\alpha} < 1$), whenever real balances (output) are below the target this policy recommends to reduce the money supply below the target, since it adapts to the expected low money demand. Such a recommendation is the opposite of the recommendation under the "optimal" policy d^O ; which only takes into account the current period downturn, but expects demand to be at the target level the following period.

The REE under the I policy is characterized by the $\tilde{A}^{(1)}$ map (5): There is only one SREE corresponding to the target $\frac{1}{4}$ and there is a continuum of REE paths with the property that in the long-run money loses its value. Notice that when $\mathcal{U}^{\alpha} = 1$; I is equivalent to F. Of course, along non-stationary REE paths there is an element of irrationality on the part of the government since its inertia assumption is not satis…ed.

In summary, we consider the three alternative stabilization policies, O; F and I: However, it should be clear from our discussion that, within our class of models, other policies may be considered, re‡ecting central bank perceptions of how the private sector will forecast in‡ation given its announced policy. Nevertheless, a careful stability analysis of our benchmark policies may help us to understand how policies should be modi…ed in order to enhance stability properties. In particular, we are interested in contrasting the performance

⁶See, for example, Marimon and McGrattan (1994) and Fudenberg and Levine (1998).

of the so-called "optimal" policy with the other two policies. To do this, in what follows we describe the dynamics of the model with adaptive private agents and a linear demand

$$
m^{d}(\mathcal{V}_{t+1}^{e}) = b_{i} \quad \mathcal{V}_{t+1}^{e}:
$$
 (9)

As we will see, while the design of an optimal …scal and monetary mix, under rational expectations, does not place any restriction on b μ^{α} ; other than b $\mu^{\alpha} > 0$; the saturation value b may determine the success of the in‡ation target \mathcal{U}^{α} : The fact that the stability of the in‡ation target may be a¤ected by the point of currency collapse, even if a collapse never occurs, is a general feature of our results. Our linear demand formulation simpli…es the corresponding analysis.

2.2 Introducing …scal constraints

Non-negative prices require m_{t+1}^d i $d_{t,s}$ 0. Here, we follow E-H-M in considering constrained policies that satisfy $m^d_{t+1|i}$ $d^{(P)}(m^d_t)$, 0: In particular, we consider a constraint on the ratio of seignorage to (private) GDP,7

$$
\frac{d_t}{y_t i} \qquad (10)
$$

By equation (3),

$$
\frac{d_t}{m_{t+1}^d} = 1 + \frac{m_t^d \textcolor{blue}{\mathcal{U}_t}^1}{m_{t+1}^d} = 1 + \frac{c_t}{m_{t+1}^d} = 1 + \frac{y_t + g}{m_{t+1}^d} \cdot \ 1 + \frac{1}{\cdot} \frac{d_t}{m_{t+1}^d};
$$

that is,

$$
\frac{d_t}{m_{t+1}^d}\cdot\ \frac{\cdot}{1+\cdot}\ \stackrel{\cdot}{\cdot}\ _{\circ}:
$$

Notice that if, instead, the constraint is a de…cit to (private) GDP constraint of the form

$$
\frac{g + (R_{t i} \t 1)b_{t i} \t i n_t}{y_{t i} g} = \frac{d_t + (b_{t+1 i} \t b_t)}{y_{t i} g}.
$$

then $d_t = m_{t+1}^d$ · , as long as $(b_{t+1} \mid b_t)$, 0, as in a (targeted) steady state budget. We abstract from the exact nature of the constraint, but we assume that ex-post policies satisfy

$$
\mathbf{d}^{(P)}(m_t^d; m_{t+1}^d) = \min f d^{(P)}(m_t^d); \, \Box m_{t+1}^d g \tag{11}
$$

 7 For example, in the EMU seignorage of the ECB is restricted; furthermore the Growth and Stability Pact constraints de…cits and, in the US, balanced budget proposals are recurrently being considered.

for some policy parameter _s. In particular, we are interested in studying how the stability of in‡ation target policies is a¤ected by such a …scal constraint parameter. It should be noticed, however, that such a constraint does not mitigate (and may actually worsen) the indeterminacy problem of REE. More speci…cally, with full commitment and rational expectations there is no rational for imposing constraints of this type (see E-H-M). Of course, with limited commitment and rational expectations there may be a stabilizing role for …scal constraints (see, for example, Giovannetti, Diaz, Marimon and Teles, 2000). As in E-H-M, this paper shows that with full commitment and adaptive expectations there is also a stabilizing role for …scal constraints.

Precautionary savings

Unfortunately, the sconstraint is not enough to avoid currency collapses (i.e., it quarantees 1= p_t , 0 but not 1= $p_t > 0$). One may consider policies explicitly aimed at avoiding such extreme events, however. As long as there is some minimum (residual) demand for money, currency collapses cannot occur. Here, as in E-H-M, we assume the existence of an $2 > 0$, such that the representative agent's demand for real balances satis...es: $m^{d}(\mathcal{U}^{e}) = \max f b$ i \mathcal{U}^{e} ; ^{2}q : As we will see, such an assumption will only play a role in our global analysis in the sense that, without it, the rare event of a currency collapse can not be dismissed.⁸

2.3 Introducing adaptive expectations

We consider that private agents predict in‡ation as a constant. In other words, we follow Cagan (1956) in considering a general class of learning rules where agents condition data focusing on a minimal state variable (MSV) solution. In particular,

$$
M_{t+1}^e = M_t^e + \mathcal{D}(M_{t+1} + M_t^e) \tag{12}
$$

where previous period, and not current period, in‡ation is used to update forecasts. This formulation is consistent with the underlying informational structure of the model and with agents not over-reacting to current events (i.e., having some behavioral inertia).⁹ We also assume that the weight on realized in‡ation, \mathcal{O}_{t} ; is exogenous. Nevertheless, experimental evidence shows that the parameter \mathcal{R}_t tends to increase when observed paths are non-stationary.

⁸Notice that, for notational convenience we also denote by $m^d(\mathcal{U}^e)$ the demand for real balances with precautionary savings.

⁹ Lettau and van Zandt (1999) show that, in contrast with Marcet and Sargent (1989), if agents react to current prices and do not focus on MSV solutions, the stability properties of the adaptive learning process changes. However, recently, Adam (2000) has shown that if Cagan's hyperin‡ationary model is properly developed as to meaningfully allow for conditioning on current prices most of the Marcet and Sargent results prevail.

In fact, in a non-stationary environment, to use a tracking procedure (i.e., keeping \mathcal{O}_t constant) is a better learning rule than to use a stochastic approximation procedure (with \mathcal{B}_t & 0), such as standard least squares procedures. Since, on the one hand, the asymptotic analysis of the stochastic approximation case has been done by E-H-M (only for the F policy) and, on the other hand, we want to allow for a wide range of tracking procedures, we should consider the whole class $\mathcal{B}_t = \mathcal{B}$ 2 $(0, 1)$:¹⁰

3 The dynamic model with adaptive expectations

In this section we provide the main stability results. We start by considering some general properties of the adaptive expectations process under a general in‡ation target policy $d^{(P)}$: Given such a policy, substituting (12) into the intertemporal equilibrium condition (4), we obtain a second order di¤erence equation in expected in‡ation rates:

$$
M_{t+1}^{e} = (1 \mathbf{i} \otimes M_{t}^{e} + \mathbf{A}^{(P)}(M_{t_{i-1}}^{e}; M_{t}^{e});
$$

which, under our assumptions, takes the form

$$
W_{t+1}^{e} = (1 \text{ i } \circledast) W_{t}^{e} + \circledast \frac{\max b_{i} W_{t_{i}+1}^{e}}{\max fb_{i} W_{t}^{e} \cdot 2g_{i} \frac{d^{(p)}(W_{t_{i}+1}^{e})}{d^{(p)}(W_{t_{i}+1}^{e})}.
$$
 (13)

As usual, a second order di¤erence equation is more easily studied by writing it as an equivalent system of two …rst order di¤erence equations. In order to do this, let ${\sf x_t}=\tt{\mathcal{U}}^e_{t_{i-1}}$ and $y_t = V_t^e$. Then equation (13) can be written in the form $(x_{t+1}; y_{t+1}) = T^{(P)}(x_t; y_t)$, where $T^{(P)}$ is the two-dimensional map

$$
T^{(P)}: \begin{cases} x_{t+1} = y_t \\ y_{t+1} = (1 + \Phi) y_t + \Phi \frac{m(x_t)}{m(y_t)_i} \frac{m(x_t)}{d^{(P)}(x_t, y_t)} \end{cases}
$$
 (14)

with $m(z) = \max f^2$; b i zg, $d^{(p)}$ given by equation (11), b $\frac{1}{2}$ $\frac{1}{4}$ $\frac{1}{8}$ $\frac{1}{2}$ $(0; 1)$, $\frac{1}{2}$ $(0; 1)$ and $2 > 0$ is a small parameter.

Some general properties of the models

The map (14), whose iteration de…nes the time evolution of the system in the space of expected in‡ation, is a nonlinear piecewise continuous map on R_+^2 . However, its behavior

 10 Notice that one could also consider that agents give some weight to the announced target, such as, $\mathcal{H}_{t+1}^e = (1 - \epsilon_t) \left[(1 - \epsilon_t) \mathcal{H}_t^e + \mathcal{H}_t \mathcal{H}_{t+1} + \epsilon_t \mathcal{H}_t^e \right]$. But, while such rule will tend to help the stability properties of the target, it complicates the analysis without providing new insights.

changes along the lines $x = b_2$ and $y = b_2$, where $b_2 \rvert b_1$ ². Correspondingly, we can subdivide R^2_+ in the following four regions:

$$
R(I) = f(x; y)j0 \cdot x < b_2; 0 \cdot y < b_2g
$$

\n
$$
R(II) = f(x; y)jx \cdot b_2; 0 \cdot y < b_2g
$$

\n
$$
R(III) = f(x; y)jx > b_2; y > b_2g
$$

\n
$$
R(IV) = f(x; y)j0 \cdot x < b_2; y > b_2g
$$

Notice that, by assumption, E^{\prime} (μ^{α} ; μ^{α}) is in region R(I) and outside this region there is only a residual, ², demand for real balances (i.e., for ${\sf m}_{\sf t_{\sf i}}^{\sf d}$ in R(II) or ${\sf m}_{\sf t}^{\sf d}$ in R(III) or both $m_{t_{i-1}}^d$ and m_t^d in R(IV)). Therefore, we are particularly interested in the behavior of (14) in R(I). The following result shows that, provided the …scal constraint is not too loose, the regions, R(II), R(III) and R(IV) are transition regions.

Lemma 1. Assume $\frac{1}{2}$ < 1 $\frac{1}{2}$ l =b₂: Then, for any initial condition (x_0, y_0) 2 R^2 a process f_{x_t} ; y_t q generated by (14) visits R(I) in...nitely often. In particular, either R(I) is an absorbing region for f_{x_t} ; $y_t q$ or, eventually, f_{x_t} ; $y_t q$ follows a path through the regions $R(I)!$ R(IV)! $R(III)!$ R(II)! R(I).

Proof: (see Appendix A)

Notice that when the …scal constraint is binding the map(14) reduces to the (sub)map

$$
T_{.}: \quad \begin{array}{l} x_{t+1} = y_t \\ y_{t+1} = (1 \, \mathbf{i} \, \mathbf{e}) y_t + \frac{\mathbf{e}}{1 \, \mathbf{i} \, \mathbf{e}} \frac{m(x_t)}{m(y_t)} \end{array} \tag{15}
$$

which has a unique …xed point at $E_{\frac{1}{2}} \leq \frac{1}{1} \frac{1}{1+\frac{1}{2}}$; $\frac{1}{1+\frac{1}{2}}$ ¢ . The assumption of Lemma 1 implies that E_i is in region R(I), that is, the condition \sqrt{s} < 1 i 1=b₂ guarantees that the map T_i, active when there is only a residual demand for real balances, does not allow the process to be absorbed outside region (I). It does not guarantee, however, that the process eventually remains in region (I) since there may be cycling behavior along the four regions. In fact, Lemma 1 allows the existence of cyclic dynamics, which may be periodic or not, that move "clockwise" visiting the four regions in the order $R(I)! R(IV)! R(III)! R(II)! R(I)$, with fast transitions (just one time period) for $R(II)! R(I)$ and $R(IV)! R(III)$ or $R(II)$, and with slower transitions for $R(III)$! $R(II)$ and $R(I)$! $R(IV)$. The existence of this type of large amplitude oscillation is strictly related to the value of the parameter 2 , in the sense that the amplitude of the oscillations is inversely proportional to ². We return to this issue when we analyze the global dynamics of the models.

3.1 Local stability of \mathbb{V}^n

We …rst study the asymptotic stability (i.e., whether \mathcal{W}_t^e $_1$! \mathcal{W}^a) of paths with initial conditions in a neighborhood of the target (i.e., $k(\frac{1}{6}; \frac{1}{6})$ i Ek < ½ for some ½ > 0). Such local stability analysis of (14) around \mathcal{U}^{μ} is relatively straightforward. It requires the characterization of the map (14) in region (I), possibly establishing conditions guaranteeing that the …scal constraint is not binding for expectations close to the target and, …nally, studying the eigenvalues of the corresponding Jacobian. We …rst brie‡y discuss the three policies and then compare them in terms of their local stability properties. For all the policies, in the subregion of R(I) where the …scal constraint is not binding the map (14) reduces to the (sub)map $T_{n}^{(P)}$ whose …xed points are the same than those of the rational expectations map (5). For convenience, policies will be discussed in reverse order with respect to their appearance, that is: I ; F and O. In what follows, given an in‡ation target \mathcal{U}^{α} , we let Ω f(b; ®) jb > ¼ⁿ; ® 2 (0; 1)g:

3.1.1 Policy I

The restriction of T^(I) to region R(I) is given by

$$
T^{(1)}j_{R(1)}: \quad x^0 = y
$$

\n
$$
y^0 = (1 \, i \, \otimes y) + \otimes \frac{m(x)}{m(y) \, i \, \min \{ \frac{y_1 \, i \, \cdot \, 1}{y_4 \, i \, \cdot \, 1} m(x) \, ; \, \frac{m(y)}{y} \} }
$$
\n(16)

The line s, of equation $y = s(x) \frac{y_a^{\alpha}+1}{y_a^{\alpha}}x + \frac{1}{1}i \frac{y_a^{\alpha}+1}{y_a^{\alpha}}$ b₂ separates the R(I) into two subregions

$$
R(I_A) = f(x; y) 2 R(I)j y < s(x)g \text{ and } R(I_B) = f(x; y) 2 R(I)j y > s(x)g
$$

The map $T^{(1)}j_{R(1)}$ can be written in the equivalent form

$$
\begin{array}{ccc} \mathbb{S} & \text{if } (x_t;y_t) \ 2 \ R(I_A) \\ \text{if } (x_t;y_t) \ 2 \ R(I_B) \\ \text{if } (x_t;y_t) \ 2 \ R(I_B) \\ \text{if } (x_t;y_t) \ 2 \ R(I_B) \end{array}
$$

The map $T_x^{(1)}$ has the unique …xed point $E^a = (4^a; 4^a)$, which is also a …xed point of T⁽¹⁾ provided that E^{π} 2(I_A), that is, if the condition $1 \nvert 1 = 4^{\pi} < \sqrt{1 - 1}$ holds. In other words, the target equilibrium is a steady state of the model if the ...scal constraint on seignorage is not too tight. With such a condition the …xed point of the map T_{g} , $E_{\text{g}} = \frac{1}{T_{\text{H}} \cdot 2}$; $\frac{1}{T_{\text{H}} \cdot 2}$ $\check{}$; is not a …xed point of T^(I), i.e., T^(I)(E_₂) = T^(I)(E₂) 6 E ₂.

We will restrict the ...scal constraint to satisfy

$$
2 \Delta^{\mathfrak{n}} \upharpoonright (1 + 1 = 4^{\mathfrak{n}}; 1 + 1 = b_2): \tag{17}
$$

As long as condition (17) is satis...ed, \Box does not a¤ect the local stability properties of Eⁿ. Indeed, let $\Omega_s^1 = f(b; \mathcal{B}) \geq \Omega_b^1 > \mathcal{H}^{\alpha}(1 + \mathcal{B}\mathcal{H}^{\alpha})g$ (see the region below the line OD in ...g. 1). The following result is proved in the Appendix B.1

Lemma 2. Assume \Box 2 Δ^{α} (i.e., condition (17)). If (b; ®) 2 Ω_{s}^{1} , then E^{α} is locally stable with policy I: In the complementary region $\Omega_u^1 = \left(b; \Phi \right)$ jb $\frac{1}{2}$ $\frac{1}{4}$ $\frac{1}{4}$ following the arguments given in the Appendix B.1, if the point (b, \mathcal{O}) crosses the line

$$
\mathbf{b} = \mathbf{b}_{\mathsf{h}}^{(1)}(\mathbf{B}) = (1 + \mathbf{B}\mathbf{W}^{\mathsf{h}})\mathbf{W}^{\mathsf{h}} \tag{18}
$$

passing from $\Omega_{{\sf u}}^{\rm I}$ to $\Omega_{{\sf s}}^{\rm I}$; a subcritical Neimark-Hopf bifurcation occurs which, at least for (b; ®) 2 Ω_s^1 close to the bifurcation curve (18), creates a repelling closed invariant curve Γ around the stable …xed point E^{α} , which constitutes the boundary of the basin of attraction $B(E^{\pi})$ of E^{π} . More precisely, for $b > b_h^{(1)}(\mathcal{R})$ a range of values of b exists such that E^{π} is locally asymptotically stable (a stable focus), with a basin of attraction bounded by a closed curve whose radius is proportional to $\mathcal{A}_{\mathcal{I}}$ b $i \; b_n^{(1)}(\mathcal{R})$, and, analogously, for a …xed value of the parameter b 2 (\mathcal{U}^{α} ; \mathcal{U}^{α} (1 + \mathcal{U}^{α})), the subcritical Neimark-Hopf bifurcation occurs at

$$
^{\circledR} = \,^{\circledR}_{h}(b) = \frac{b \, \, j \, \, \mathcal{U}^{n}}{\mathcal{U}^{n^{2}}}
$$
 (19)

and E[¤] is stable for ® < ® $_{{\sf h}}^{({\sf I})}$ with basin of attraction bounded, at least for values of ® close to $\binom{\mathfrak{G}(1)}{\mathfrak{h}}$, by a closed curve whose radius increases proportionally to $\sqrt{\frac{\mathcal{B}(1)}{\mathcal{B}(1)}(b)}$ i \mathcal{B} .

3.1.2 Policy F

The restriction of $T^{(F)}$ to the region (I) is given by

$$
T^{(F)}j_{R(1)}: \begin{cases} x_{t+1} = y_t \\ y_{t+1} = (1 \, \mathbf{i} \, \mathbf{e})y_t + \mathbf{e} \frac{m(x_t)}{m(y_t)_i \, \min \{ \frac{M^R + 1}{M^R} m(W^R)_i \cdot m(y_t) \}} \end{cases} \tag{20}
$$

The horizontal line q, of equation $y = q(x)$ $\leq b_2$ $\frac{y_a^{\alpha+1}}{x^{\alpha}}$ $(b_2 \neq y_a^{\alpha})$ separates the region R(I) into two subregions

$$
R(I_A) = f(x; y) 2 R(I)j y < q(x)g \text{ and } R(I_B) = f(x; y) 2 R(I)j y > q(x)g
$$

such that the map T $^{\rm{(F)}}$ j_(I) can be written in the equivalent form

$$
\begin{array}{ccc} \mathbb{S} & \pi^{(F)}j_{(I_{B})} = T \, : & \text{if} & (x_{t}; y_{t}) \, 2 \, R(I_{A}) \\ \mathbb{S} & \pi^{(F)}j_{(I_{A})} = T_{a}^{(F)}: & \begin{array}{cc} x_{t+1} = y_{t} & & \\ y_{t+1} = (1 \, i \, \end{array} \end{array}
$$

As the corresponding REE map (5), the map $T_{\alpha}^{(F)}$ has two ... xed points: the target E^{π} = $(\mathcal{H}^{\alpha}; \mathcal{H}^{\alpha})$, and $B^{\alpha} = (b - \mathcal{H}^{\alpha}; b - \mathcal{H}^{\alpha})$. These points are also …xed points for $T^{(F)}$ provided that they belong to the region (I_A) where the dynamics of $T^{(F)}$ are governed by the restriction $T_{n}^{(F)}$. It is easy to see that $E^{n} 2(I_{A})$ if $\sqrt{s} > 1$; $\frac{1}{\sqrt{a}^{n}}$ (which is satis...ed if condition (17) holds) and B^{_{¤}} 2(l_A) if _> > 1 _i $\frac{1/4}{b}$. (Notice that 1 _i $\frac{1}{1/4}$ < 1 _i $\frac{1/4}{b}$ if b > 1/4^{n}2). As with policy 1; we</sup> assume that condition (17) is satis…ed.

On the basis of the analysis of the eigenvalues given in the Appendix B.2, the target ... xed point E^{π} is stable in the region $\Omega_S^F = (b; \mathcal{B}) 2 \Omega/b > 4^{\pi^2}$ and $b > 4^{\pi}(1 + \mathcal{B})$ (see the shaded region bounded by the lines AB and BC in ...g. 1) and for (b; Φ) 2 Ω_s^F B^{π} is a saddle point. The two …xed points of $T_{\text{H}}^{(F)}$ exchange stability via a transcritical bifurcation at b $=$ 14 $^{\mathfrak n^2}$ at which E $^{\mathfrak n}$ $=$ B $^{\mathfrak n}$, so that the …xed point characterized by lower in‡ation is the stable one.

The unique …xed point E_{\perp} of the map T_{\perp} is also a …xed point for $T^{(F)}$, provided it belongs to the region (I_B) , i.e. b_2 _i $\frac{V_A^H + 1}{2} (b_2 + V_A^{\alpha}) < \frac{1}{1 + \alpha} < b_2$. Furthermore, E_s is locally stable provided b $> b_{\hat{h}}(\textcircled{\tiny{\textcircled{\tiny{\textcirc}}}}) = \frac{\textcircled{\tiny{\textcircled{\tiny{\textcircled{\tiny{\textcirc}}}}}+1}{1_{i}}$ (see the Appendix B.4). From these conditions for the existence and stability of the …xed points, we obtain the following result:

Lemma 3. Assume $\frac{1}{2}$ Δ^{α} (i.e., condition (17)) and let (b; ®) 2 Ω

- (i) If $b < \lambda^{n^2}$ then the map $T^{(F)}$ has three ... xed points: E^{α} , B^{α} and E_{α} . If $\omega \in \lambda^{\alpha}$ i 1 then E^{π} is unstable and B^{π} is stable, while if \mathcal{P} < $(1 + \sqrt{2})$ i 1 then E_{π} is locally stable ;
- (ii) The target E^{π} is locally stable provided that $(b; \mathcal{B}) \geq \Omega_s^F$. Furthermore, if $\zeta < 1$ i $\frac{W^{\pi}}{D\sigma}$ then E^{π} is the only ... xed point of $T^{(F)}$; while if $1 + \frac{1}{2}$ i $\frac{1}{2}$ i $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $1 - 1$ = $T^{(F)}$ has three …xed points, E^{α} , B^{α} and E_{α} , where B^{α} is unstable, and E_{α} is stable if \mathbb{B} < $(1 \, \text{i} \, \text{j})$ b i 1:

As with policy I, for b > \AA^{u^2} and high values of ® the target equilibrium \AA^{u} is unstable (the only attractor being a big "cyclic" set $A(2)$), then E^{α} becomes stable for decreasing values of ® through a subcritical Neimark-Hopf bifurcation at the line $b = b_h^{(F)}(\mathcal{B}) = \frac{\mu^{\pi}(\mathcal{B} + 1)}{2}$:

3.1.3 Policy O

The restriction of $T^{(O)}$ to region (I) is given by

$$
T^{(O)}j_{(1)}: \begin{cases} x_{t+1} = y_t \\ y_{t+1} = (1 \, i \, \Phi)y_t + \Phi \frac{m(x_t)}{m(y_t)_i \, min \, \text{Im}(x_t)_i \, \frac{1}{\gamma_t^2} m(x_t)_i \, \frac{m(y_t)}{g}} \end{cases} \tag{21}
$$

The line r, of equation $y = r(x)$ $\frac{1}{i} \frac{1}{\sqrt{4^n}} x + \frac{1}{2} \frac{1}{i} \frac{1}{\sqrt{4^n}} + \frac{1}{2} \frac{1}{i} \frac{1}{i}$ separates the region R(I) into two subregions

$$
R(I_A) = f(x; y) 2 R(I)j y < r(x)g \text{ and } R(I_B) = f(x; y) 2 R(I)j y > r(x)g
$$

such that the map T $^{(O)}j_{(1)}$ can be written in the equivalent form

$$
\begin{array}{ll} \mathbb{S} & \text{if } (x_t; y_t) \ 2 \ R(I_A) \\ \geqslant & T^{(O)}j_{(I_B)} = T \\ \text{if } (x_t; y_t) \ 2 \ R(I_A) \\ \geqslant & T^{(O)}j_{(I_A)} = T^{(O)}_{{}_H}: \ \ \begin{array}{ll} x_{t+1} = y_t \\ y_{t+1} = (1_{\,\,\, \text{if } \ \mathbb{O}}) y_t + \frac{\Theta y_t^{\alpha}}{m(y_t)_{\,\, \text{if } \ (W_t^{\alpha})_i \ \frac{1}{y_t^{\, \, \text{if } \ (X_t; Y_t) \ 2}} \ R(I_B) \end{array} \end{array}
$$

As the corresponding REE map (5), $T_a^{(O)}$ has two …xed points: the target $E^a = (\mathcal{V}^a; \mathcal{V}^a)$ and $B^{\alpha} = (b=(1 + \lambda^{\alpha})$; $b=(1 + \lambda^{\alpha})$. These points are also ... xed points for T^(O) provided that they belong to the region (I_A) where the dynamics of $T^{(O)}$ are governed by the (sub)map $T_n^{(O)}$. It is easy to see that, with condition, (17) E^{π} 2R(I_A) and B^{π} 2R(I_A) if $\frac{1+i\lambda^{\pi}}{b}$. (Notice that $1_i \frac{1}{\mu^{\pi}} < 1_i \frac{1+\mu^{\pi}}{b}$ if $b > \mu^{\pi}(1+\mu^{\pi})$). Therefore, the characterization is similar to that obtained for policy F. With policy O; the two ... xed points of $T_{\text{N}}^{(\text{O})}$ exchange stability via a transcritical bifurcation at $\mathsf{b}=\mathsf{M}^{\mathsf{u}}(1+\mathsf{M}^{\mathsf{u}})\mathsf{b}=\mathsf{M}^{\mathsf{u}^2}$ at which $\mathsf{E}^{\mathsf{u}}=\mathsf{B}^{\mathsf{u}}.$ As with policy O the …xed point characterized by lower in‡ation is locally stable under adaptive learning. However, in contrast to it, in this case the condition $b > \frac{1}{4}$ $(1 + \frac{1}{4}a^{\frac{1}{2}})$ is the only condition for the stability of E[¤], that is $\Omega_s^O = f(b; \Phi)$ 2 Ω jb > ¼ $^u(1 + V_a^u)$ g (see the shaded region on the right of the line ED in …g. 1). The unique …xed point E_{\perp} of the map T_{\perp} is also a …xed point for $T^{(O)}$ provided it belongs to the region R(I_B); that is, if $\frac{1}{\sqrt{\lambda^n}}$ $\frac{1}{1} + \frac{y_a}{1} + \frac{b^2}{2} + \frac{1}{y_a} + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \frac{1}{5} + \frac{1}{5} + \frac{1}{6} + \frac{1}{7} + \frac{1}{8} + \frac{1}{9} + \frac{1}{1} +$ Furthermore, E_s is locally stable provided that $b > \frac{\circledast + 1}{1}$ (see the Appendix B.4). In summary, we obtain local stability results that almost parallel those of policy F:

Lemma 4. Assume $\frac{1}{2}$ Δ^{α} (i.e., condition (17)) and let (b; ®) 2 Ω

(i) If $b < \frac{u}{a}(1+\frac{u}{a})$ then the map $T^{(O)}$ has three …xed points, E^{α} , B^{α} and E . E^{α} is locally unstable, B^{α} is locally stable, while E is locally stable if $\mathcal{B} < (1 \, \text{i})$, β i 1;

(ii) The target E^{π} is locally stable provided that (b; ®) 2 Ω_s^O . Furthermore, if ζ < 1 ζ (1 + \mathcal{U}^{α})=b₂ then E^{α} is the only …xed point of T^(O); while if 1 _i $(1 + \mathcal{V}^{\alpha}) = b$ ₂ < ₂ < 1 _i 1=b₂ then the map $T^{(O)}$ has three ... xed points, E^{π} , B^{π} and E , where B^{π} is unstable, and E is stable if $\mathcal{B} < (1 \text{ i } \text{ s})$ b i 1:

3.1.4 Ranking policies according to their local stability properties

Lemmas 2-4 show how the local stability properties of the in‡ation target \mathcal{U}^{α} di¤er across policies. In particular, assuming condition (17), the stability of the in‡ation target under the policies I, F and O holds in the following domains of the parameters' space Ω

$$
\Omega_{s}^{l} = f_{l}^{(b; \circledast)} 2 \Omega_{l}^{j} b > \frac{1}{4} \Omega_{s}^{m} (1 + \Omega_{s}^{m}) g
$$
\n
$$
\Omega_{s}^{F} = (b; \circledast) 2 \Omega_{l}^{j} b > \frac{1}{4} \Omega_{s}^{m} (1 + \Omega_{s}^{m}) g
$$
\n
$$
\Omega_{s}^{O} = f_{l}^{(b; \circledast)} 2 \Omega_{l}^{j} b > \frac{1}{4} \Omega_{s}^{m} (1 + \frac{1}{4} \Omega_{s}^{m}) g
$$

Therefore, we say that policy P dominates policy P', in terms of its local stability properties, if the in‡ation target equilibrium $\frac{1}{4}$ is locally stable in a larger domain of the parameter space Ω , and denote such preference by P \hat{A}_1 P', then as corollary to Lemmas 2-4 we have

Proposition 1. Assume $\frac{1}{2}$ Δ^{α} (i.e., condition (17)) and let $\mu^{\alpha} > 1$: Then policy O is dominated in terms of its local stability properties. In particular, $F \hat{A}$ O and I \hat{A} O:

Figure 1 illustrates Proposition 1 for $2 > \frac{1}{4}$ > 1: Notice that, as long as $b > \frac{1}{4}$, the unconditional policy F dominates the other policies in terms of its local stability properties. This result is consistent with Friedman's views.

A local stability ranking is not uniquely determined by \hat{A}_1 : For example, provided that the target is locally stable, we may be interested in whether convergence is monotone, which can make it easier to "pattern recognize" the tendency for in‡ation to converge to the target. Alternatively, we may be interested in the speed of convergence to the target. As we show in Appendix C, provided that the target is locally stable, only with policy O convergence is always monotone, while for other policies monotone convergence requires a small enough value of ®: Nevertheless, in terms of speed of convergence, policy O also tends to be dominated.

More formally, let $\Omega_s^{\tt n} = \Omega_s^{\tt l} \setminus \Omega_s^{\tt F} \setminus \Omega_s^{\tt O}$, i.e., $\Omega_s^{\tt n}$ ½ Ω denote the region of parameters where the in‡ation target is locally stable under the three policies under consideration. We say that P \hat{A}_s P' on a subset A μ $\Omega_s^{\tt w}$ if for any $({\tt b}; \circledast)$ 2 A paths (starting in a neighborhood of

 \mathcal{U}^{α}) converge faster under the policy P than under the policy P'. The following proposition (proved in the Appendix C) provides the corresponding characterization.

Proposition 2. Assume $(b; \mathcal{D})$ 2 Ω_s^{π} and $\mathcal{U}^{\pi} > 1$: Then

- (i) there exists an Φ such that, for all $\Phi \cdot \Phi$ all three policies have monotone path,
- (ii) there exists an \mathcal{B}_1 2 $(\mathcal{B}_1, 1)$ and an \mathcal{B}_2 2 $(\mathcal{B}_1, 1)$ such that, for all \mathcal{B}_1 \mathcal{B}_2 \mathcal{B}_3 \mathcal{B}_4 and, for all \mathscr{B} . \mathscr{B}_2 , F \hat{A}_s O:

[INSERT FIG. 1]

As we have shown, the local stability analysis already allows us to rank in‡ation target policies, and in particular it suggests disregarding the "optimal" policy O in favor of alternative policies. On the other hand, di¤erences based on the eigenvalues of the Jacobian of $T_{n}^{(P)}$ tend to be relatively small and, therefore, the rankings are not very sharp. We now turn in the next subsection to the more interesting and novel global analysis of the three policies.

3.2 Global stability of \mathcal{U}^{α}

As in the previous subsection, we …rst brie‡y discuss global dynamics under the alternative policies and we then summarize the results comparing the three policies. As we will see, even if the local analysis also provides useful information concerning the global dynamics of the system, a more complete understanding is based on the study of the basins of attraction and, in particular, of some global bifurcations which cause qualitative changes of such basins, whose characterization requires the use of computer graphics. We focus our attention on the basin of attraction of \mathcal{U}^{α} ; B (E^{α}), de ... ned as the set of points of the plane x; y which generate trajectories converging to E^{α} . Of particular interest is the role played by the ...scal constraint parameter and by the tracking parameter \mathcal{P} in enlarging B (E^{α}). The global analysis becomes quite complex due to the possible coexistence of di¤erent attractors. As we will see, in all these respects the three, apparently very similar, policies behave quite di¤erently. Such di¤erences could not be captured in a model where only the asymptotic case $\mathbb{R} \otimes \mathbb{Q}$ is analyzed (for example, in E-H-M).

3.2.1 Policy I: the role of …scal constraints

As we have seen in Lemma 1, even when the in‡ation target is locally stable, there may be cycling paths following a large cyclical movement across the four regions. Figure 2 illustrates such behavior for policy I: In particular, Figure 2a shows, in the phase space x_i , y, the coexistence of a large "cyclic" attractor $A(2)$, whose basin is represented by the white region, with the SREE \mathcal{U}^{α} whose basin B (E^{α}) is represented by the grey region. Figure 2b shows two paths each one starting from an initial expected in‡ation taken in a di¤erent basin attraction.

[INSERT FIG. 2]

In Figure 2a, $B(E^{\pi})$ is contained in the interior of subregion (I_A) . This is a snap-shot corresponding to …xed values of b; ® and : Nevertheless, changing these parameters also causes $B(E^{\alpha})$ to change: In particular, numerical simulations show how the size of B (E^{α}) increases for decreasing values of \mathcal{P} (or increasing values of b) until the basin boundary $\mathcal{P}B(E^{\alpha})$ has a contact with the big "cyclic" attractor $A(2)$. This contact causes the disappearance of $A(2)$ (Gumowski & Mira, 1978, 1980) and consequently E^{π} becomes a global attractor, i.e. B (E^{π}) covers the whole phase space. Such a contact bifurcation is called …nal bifurcation in Mira, et al. (1996a) and Abraham, et al. (1997) or boundary crisis in Grebogi, et al. (1983). This bifurcation cannot be revealed by a local study, that is, based on the linear approximation of the dynamical system.

An interesting result is obtained if the in‡uence of the parameter ¸ on the size and the shape of B (E^{α}) is considered. In fact, even if β does not in‡uence the local stability of E^{α} when condition (17) is assumed, it may in‡uence the shape and the size of B (E^{π}): This is clearly shown in Figure 3, where we start with a situation similar to that of Figure 2a (see Figure 3a) and, keeping all the other parameters ... xed, we successively decrease, making the …scal constraint tighter. In Fig. 3b B (E^{α}) intersects the subregion (I_B) where dynamics are dominated by the (sub)map T : The contact between the basin boundary $\mathscr{B}(E^{\alpha})$ and the line s, which separates the subregions (I_A) and (I_B) , causes a sudden enlargement of the basin B (E^{r}). In fact, after such contact, if E is stable for T and E 2 B (E^{r}), then some trajectories starting from region (I_B) may move toward E and consequently enter the basin $B(E^{\alpha})$: We may say that E behaves as a catalyst, since it attracts trajectories coming from the subregion (I_B) and then it conveys them towards E^* because $E \cdot 2 B(E^*)$. Moreover, a small reduction of \Box causes B (E^{\Box}) to increase to the point where the basin boundary @B (E^{\Box})

contacts the line $b₂$ (see Fig. 3c) producing a global (or contact) bifurcation. As Fig. 3d shows, as a result of such a global bifurcation $B(E^n)$ covers the entire phase space under consideration, so that global stability is achieved.

In summary, Figure 3 shows how …scal constraints can enhance the global stability properties of an in‡ation target policy (such as I) even when the constraints have no e¤ect on local stability properties of the in‡ation policy.

It is important to remark that, since the equations of the curves which form $\mathscr{B}(E^{\alpha})$ are not known, an analytical computation of the parameters values at which the contacts between $\mathscr{B}(E^{\alpha})$ and the lines s and $b = b_{\alpha}$ occur is not possible—hence these parameters can only be revealed numerically, by a graphical analysis. Indeed, computational methods are a standard tool in the global study of dynamical systems of dimension greater than one (see e.g. Mira et al. (1996), Brock and Hommes (1997)).

[INSERT FIG. 3]

3.2.2 Policy F: the coexistence of two attracting …xed points

As Lemma 3(ii) shows, the …xed points E^* and E may coexist, both being locally stable. In this case of two coexisting attractors, the initial condition is crucial in order to forecast the long-run behavior of the system; it is therefore important to study the boundaries of the respective basins of attraction. As with policy I ; when E^{α} is the only attractor, decreasing [®] or μ or increasing b, enhances the stability of \mathcal{H}^{α} , and B (E^{α}) expands. However, when both E^{α} and E are attractors, these changes of parameters tend to enhance the stability properties of both attractors and it may well be that the e¤ect is stronger for E ; in which case B (E) will enlarge while B (E^{α}) will contract. This is shown in Figure 4, where we start in a situation where both attractors coexist, but just after the subcritical Neimark-Hopf bifurcation at which E becomes stable and, therefore, $B(E^n)$ encompasses almost all of the phase space (see Fig. 4a; notice that the Neimark-Hopf bifurcation at which E becomes stable occurs at $\mathcal{D} = (1 \mathbf{i} \mathbf{j}) \mathbf{b} \mathbf{i} \mathbf{i} = 0.25$. In Figs. 4b-d we successively reduce the tracking parameter \mathcal{R} while keeping all other parameters constant. As \mathcal{R} is decreased, B (E) enlarges and its boundary has a contact with the line q. After this contact a sudden change of $B(E_+)$ is observed, as shown in Fig. 4b. Now the boundary of the basin $B(E)$ includes the saddle point B^{α} and consequently points which are very close to E^{α} belong to B (E). Furthermore, if \circ is further decreased, B(E) continues to enlarge until a contact with the line $x = b₂$ occurs (see Fig.4c) which marks another evident qualitative change, as Fig. 4d shows.

In summary, Figure 4 shows how the presence of coexisting attractors (as may occur under policy F) can induce counterintuitive e¤ects on the stability properties of the in‡ation target $\frac{1}{4}$ ^a when parameters are changed.

[INSERT FIG. 4]

3.2.3 Policy O: the coexistence of two attracting …xed points and a chaotic attractor

Lemma 4(ii) shows that, with the policy O, the ... xed points E^* and E can coexist as attractors. However, as Figure 5 shows, the situation may be more complex: In particular, Fig. 5b shows the existence of a chaotic attractor around E : In this ... qure the dark grey and the light grey regions represent the basins of E^{α} and E respectively, whereas the points of the white region converge to the chaotic attractor. Notice that the basin $B(E)$ is formed by two disjoint portions. However, as the parameter \mathcal{O} is decreased, the chaotic attractor disappears after a contact with its basin boundary, a typical …nal bifurcation (or boundary crisis), see Fig. 5b.

In summary, Figure 5 shows that the global dynamics can be quite complex. However, decreasing ® (or increasing b) tends to simplify the dynamics of the model in favor of the attracting …xed points. As in Figure 4, however, stability may be enhanced more for E than for F^{α}

[INSERT FIG. 5]

3.2.4 Comparing policies according to their global stability properties with the help of …scal constraints.

The results on global dynamics given above are interesting but do not lead to a clear ranking of policies according to their global stability properties. In order to provide such a comparison, we restrict our attention to values of $\Box 2 \Delta^{\alpha}$ \land (1 \Box 1=14^{α}; 1 \Box 1=b₂) (i.e., where condition (17) is satis...ed) and check, by numerical computation, which values of \cdot ; b and ® produce "global convergence." More precisely, given a set of parameters $(\mathcal{O},\mathfrak{b})$; we numerically generate paths from all initial conditions $(x_0; y_0)$ taken within a …ne grid in a wide portion of the (x, y) plane, and we count how many of such paths converge to the target. Figure 6 shows the results of these computations, made for many values of $(b; \,)$ (whose values are represented on the axes) and two di¤erent values of \mathcal{O} . From Lemmas 3(ii) and 4(ii), for values

 $(\square; \mathfrak{b})$ between the curves $\square^{\mathfrak{n}}(\mathfrak{b})=1$ \mathfrak{j} $1=bz; \square^{\mathsf{F}}(\mathfrak{b})=1$ \mathfrak{j} $\parallel \mathfrak{A}^{\mathfrak{n}}=bz;$ and $\square^{\mathsf{O}}(\mathfrak{b})=1$ \mathfrak{j} $\parallel (1+\mathfrak{A}^{\mathfrak{n}})=bz;$ respectively, the attractor E^{α} may coexist with the attractor E ; while for values of $(, ; b)$ below $\binom{F}{i}$ and $\binom{O}{i}$, E^{α} is the unique attractor. In contrast, for policy I there is a unique ... xed point that can be an attractor (E is in subregion (I_A)) and this results in a better performance of this policy in terms of global stability for relatively low values of ®, but for relatively high values of \odot the target may cease to be stable and policy F may dominate policy I in terms of global stability.

[INSERT FIG. 6]

In summary, Figure 6^{11} reinforces the local stability ranking of policies. In particular, the global stability results are consistent with Propositions 1 and 2 in showing that the so-called optimal policy O tends to be outperformed, as a stabilization policy, by either the unconditional Friedman policy F or the adaptive inertia policy I when private agents form their expectations adaptively.

4 Conclusions

Stabilization policies must be judged by their stability properties. Within rational expectations equilibria such a statement is not even meaningful. It is meaningful, however, when we consider that agents may form their expectations adaptively. Experimental evidence (see, for example, Marimon and Sunder, 1993, 1994, 1995) supports this adaptive view and can provide an empirical ground for our stability results.¹² The fact that our local and global stability rankings are consistent is encouraging. In particular, our results reinforce Friedman's caution against "overly reactive" rules. Friedman had an intuition about policy lags that could apply to any model. In contrast we provide a careful stability analysis of a relatively simple model without policy lags. Even though, some lessons emerge that are likely to apply to other models. First, and foremost, the misspeci…cation that private agents have rational expectations when they do not, may lead to a wrong policy design, in the sense that alternative designs of stabilization policies may outperform the rules designed under the rational expectations assumption. Second, even leaving aside time-consistency considerations or "Fiscal Theory of Money" considerations (see, for example, Woodford 1996), …scal

¹¹ Similar computations, not reported here, are available on request.

 12 In fact, Evans Honkapohja and Marimon (2000) provide some experimental results showing the stabilization power of …scal constraints.

constraints (in particular, seignorage constraints) may play an important role in helping stabilization policies to achieve their goals¹³. Third, even if monetary authorities follow—with full commitment—their announced in‡ation target rules, in‡ation may di¤er substantially from the target. While, for example, in‡ationary episodes above the target are usually associated with loose monetary policy or weak monetary authorities, in our economies such instability may well correspond to the fact that, due to the existence of money substitutes, the in‡ation target may not be too far from the level of in‡ation in which there is a currency collapse. Furthermore, our global analysis also provides a good reason to study the point of currency collapse: It is the point where a global contact bifurcation occurs, resulting in a qualitative improvement of the stability properties of the policy.

There is room for further research in several directions: studying other misspeci…ed models, introducing stochastic learning, and so on. In such extensions, it would be interesting to see if the relatively good performance (as a stabilization policy) of Friedman's constant money growth rule persists. We …nd it a remarkable result that may generalize to other environments.

A Proof of Lemma 1

We ...rst prove that all the trajectories starting out of region R(I) enter region R(I) after a …nite number of steps. In fact

(a) if $(x_t; y_t)$ 2R(II) then $(x_{t+1}; y_{t+1})$ 2R(I), because in the map (14) $y_t < b$ implies x_{t+1} < b₂ and $y_{t+1} = (1 \, \mathrm{i} \, \mathrm{e}^0) y_t$ < b₂.

(b) if $(x_t; y_t)$ 2R(III) and ζ < 1 i 1=b₂ then $(x_{t+k}; y_{t+k})$ 2R(II) for a …nite k > 0. In fact, in region R(III) we have $m(x) = 2$ and $m(y) = 2$, hence the map $T^{(P)}$ becomes

$$
T^{(P)}j_{R(111)}: \begin{array}{c} y_2 \\ x_{t+1} = y_t \\ y_{t+1} = (1 \, i \, \Phi)y_t + \Phi \frac{1}{1 \, i \, .} \end{array}
$$

This is a linear map with a triangular structure, the second component only being dependent on the second variable, and it is immediate to see that y_t converges to $1=(1 \quad \circ \quad)$ at a speed $(1 \, \mu \otimes)^t$, hence the entrance inside the region R(II) after a ... nite number of steps is ensured if $1=(1 \quad \ldots) < b_2$, i.e. $\ldots < 1 \quad \text{i} = b_2$.

(c) if $(x_t; y_t)$ 2R(IV) then $(x_{t+1}; y_{t+1})$ 2R(III) or $(x_{t+1}; y_{t+1})$ 2R(II), because $y_t > b_2$ implies $x_{t+1} = y_t > b_2$.

 13 Notice, however, that if fiscal constraints are too tight the target may no be a stationary equilibrium.

To complete the proof we now show that a trajectory may transit from region R(I) to region $R(IV)$, so that $R(1)$ is not trapping. In fact, in region $R(1)$ we have

$$
T^{(P)}j_{R(I)}: \, \, \begin{array}{l} x_{t+1} = y_t \\ y_{t+1} = (1 \, \, \text{\small i} \, \, \, \text{\small \mathfrak{D}_j} \, \, \text{\small $y_{t+1} = (1 \, \, \text{\small j} \, \, \, \text{\small j} \,
$$

from which it is evident that a movement from region (I) to region (II) is impossible, because y_t < b₂) $x_{t+1} = y_t$ < b₂, whereas we may have (x_t, y_t) 2R(1) and (x_{t+1}, y_{t+1}) 2R(IV) whenever y_t is su¢ciently close to b and x_t is su¢ciently small.¥

B Local stability analysis

In this appendix we analyze the local stability of the …xed points of the maps $T_{\text{H}}^{(1)}$, $T_{\text{H}}^{(F)}$, $T_x^(O)$ and T . Such analysis is obtained by the standard study of the eigenvalues, i.e. the solutions of the characteristic equation

$$
P(z) = z2 i Tr \t z + Det = 0 ; \t(22)
$$

where Tr and Det are the trace and the determinant of the Jacobian matrix computed at the …xed point. A su¢cient condition for the stability is expressed by the following system of inequalities

$$
P(1) = 1 \text{ i } Tr + Det > 0; \ P(\text{i } 1) = 1 + Tr + Det > 0; \ 1 \text{ i } Det > 0 \tag{23}
$$

that give necessary and su¢cient conditions for the two eigenvalues of (22) be inside the unit circle of the complex plane (see, for example, Gumowski and Mira (1980) p. 159).

B.1 Map $T_x^{(1)}$

The Jacobian matrix of the map $T_{n}^{(1)}$ evaluated at the unique …xed point E^{n} , is

$$
DT_{n}^{(1)}(\mathcal{U}_{n}^{n};\mathcal{U}_{n}^{n}) = \begin{array}{cc} 0 & 1 \\ \frac{\otimes \mathcal{U}_{n}^{n}}{b_{i} \mathcal{U}_{n}^{n}} & 1 \, \mathfrak{g} & \frac{\otimes \mathcal{U}_{n}^{n}}{b_{i} \mathcal{U}_{n}^{n}} \end{array}.
$$
 (24)

The characteristic equation (22) has coe¢cients $Tr = Tr^{(1)} = 1$ i $\mathcal{D} + \frac{\mathcal{D}V^{n^2}}{b_1 V^{n^2}}$ and Det = $Det^{(1)} = \frac{\omega_{\mathcal{H}}^{n2}}{b_1 \mu^n}$. The conditions $P(1) > 0$ and $P(j|1) > 0$ are always satis...ed, and the only condition for the stability of E^{α} is 1; Det > 0, i.e.

$$
\frac{\text{d} \mathcal{U}^{\text{m2}} \text{i } b + \mathcal{V}^{\text{m}}}{b \text{i } \mathcal{V}^{\text{m}}} < 0 \text{ :}
$$

Since $b>Y^{\alpha}$ in the parameter space Ω , a su¢cient condition for the stability of E^{*n*} is

$$
\mathfrak{B\!}_{\mathfrak{p}}^{\mathfrak{n}^2} \quad \text{if} \quad \mathfrak{b} + \mathfrak{h}^{\mathfrak{n}} < 0 \tag{25}
$$

The vanishing of the left hand side of (25) gives a line, in the parameter space b; ®, such that if $(b; \mathcal{D})$ crosses that line from left to right a pair of complex conjugate eigenvalues enters the unit circle and a subcritical Neimark-Hopf bifurcation occurs at which the …xed point E^{α} is changed from unstable focus to stable focus, and a repelling closed invariant orbit is created around i t¹⁴ (see, for example, Guckenheimer and Holmes (1983) p.162). Just after its creation, such a closed curve is smooth and approximately of circular shape, with radius proportional to the square root of the distance of the point $(b; \mathcal{O})$ from the bifurcation line, at least for values of $(b; \mathcal{B})$ close to the bifurcation curve (see e.g. Guckenheimer and Holmes (1983) p.305).

B.2 Map $T_{\text{m}}^{\text{(F)}}$

The Jacobian matrix at the ... xed point E^{α} , is

$$
DT^{(F)}(\mathcal{U}^{a};\mathcal{U}^{a})=\begin{array}{cc}0&1\\ \frac{\circ\mathcal{U}^{a}}{b_{i}\mathcal{U}^{a}}&1\mathcal{U}^{a}+\frac{\circ\mathcal{U}^{a2}}{b_{i}\mathcal{U}^{a}}\end{array}.
$$
 (26)

Hence the characteristic equation (22) has coe¢cients $Tr = Tr^{(F)} = 1$ i $\Phi + \frac{\circledast \pi^{n}2}{b_1 \cdot \pi^n}$ and Det = Det^(F) = $\frac{\otimes \mu^{\alpha}}{b_i \mu^{\alpha}}$. In this case we have P(1) = $\frac{\otimes \frac{b_i}{b_i} \mu^{\alpha} }{b_i \mu^{\alpha}} > 0$ if b > μ^{α^2} (being b > μ^{α} in the parameter space Ω). At b = μ^{n^2} the …xed point Eⁿ merges with the other …xed point Bⁿ and one eigenvalue is equal to 1 . This situation corresponds to a transcritical (or stability exchange) bifurcation. The other two conditions, $P(i|1) > 0$ and $1|i$ Det > 0 , become, respectively

$$
\frac{\mathfrak{b}(2\mathfrak{i}^{\mathfrak{B}})+\mathfrak{A}^{\mathfrak{n}^{2}}((\mathfrak{A}^{\mathfrak{n}}+2)^{\mathfrak{B}}\mathfrak{i} \ 2)}{\mathfrak{b}^{\mathfrak{i}}\ \mathfrak{A}^{\mathfrak{n}}}>0 \text{ and } \frac{\mathfrak{A}^{\mathfrak{n}}(\mathfrak{B}+1)\mathfrak{i} \ \mathfrak{b}}{\mathfrak{b}^{\mathfrak{i}}\ \mathfrak{A}^{\mathfrak{n}}}<0 \tag{27}
$$

The former is always satis...ed for (b; ®) 2 Ω , whereas the vanishing of the numerator of the latter gives a bifurcation curve at which a subcritical Neimark-Hopf bifurcation occurs.

The Jacobian matrix of the map $T_{\text{F}}^{(F)}$, evaluated in the other …xed point B^{F}, becomes

$$
DT^{(F)}(\frac{b}{\gamma_4^{n}};\frac{b}{\gamma_4^{n}})=\begin{array}{cc}0&1&\\ \frac{\circledast}{\gamma_4^{n}+1}&1&\circledast+\frac{\circledast}{\gamma_4^{n}(\gamma_4^{n}+1)}\end{array}.
$$
 (28)

¹⁴The rigorous proof of the subcritical nature of the Hopf bifurcation requires the evaluation of some long expressions involving derivatives of the map up to order three. In this case we claim numerical evidence.

In this case $P(1) = \frac{\omega_{\mu}^{n} \cdot \omega_{i}}{\omega_{i}^{n} (\mu_{i}^{n} + 1)} > 0$ if $b < \mu^{n^{2}}$. This con... rms that the stability properties of E^{π} and B^{π} are exchanged at $b = \frac{1}{4}a^{2}$, when the two ... xed points merge. The other conditions $P(i|1) > 0$ and $1|i|$ Det > 0 , become, respectively

$$
\frac{\mu^{\alpha^2}(2\mathbf{i} \otimes) + 2(\mathbb{B}\mathbf{i} \cdot 1)\mu^{\alpha} + \mathbb{B}\mathbf{b}}{\mu^{\alpha}(\mu^{\alpha}\mathbf{i} \cdot 1)} > 0 \quad \text{and} \quad \frac{\mathbb{B}\mathbf{i} \cdot \mu^{\alpha} + 1}{\mu^{\alpha}\mathbf{i} \cdot 1} < 0: \tag{29}
$$

For $\mathcal{U}^{\mu} > 1$ the …rst condition is satis…ed for each ® 2 (0; 1), whereas the second condition is satis…ed for ® < ¼ $\frac{1}{4}$ i 1. Hence, if $1 < \frac{1}{4}$ < 2 and b < ¼ $\frac{1}{4}$ the equation ® $= \frac{1}{4}$ i 1 de…nes a bifurcation curve at which a subcritical Hopf bifurcation occurs, the ... xed point B^{π} being a stable focus for $\Phi < \mu^2$ i 1. If b $> \mu^2$ then B^{π} is a saddle-point, with eigenvalues $0 < z_1 < 1$ and $z_2 > 1$, a straightforward consequence of the inequalities $P(i, 1) > 0$, $P(1) < 0$ and $P(0) > 0$. These arguments allow us to give the following classi-cation of the stability properties as the parameters \mathcal{U}^{α} , b and ® vary: If $\mathcal{U}^{\alpha} > 1$, then E^{α} is a locally stable ... xed point if

$$
b > \mu^{n^2}
$$
 and $b > b_h^{(F)}(\mathcal{R})$; with $b_h^{(F)}(\mathcal{R}) = \mu^{n}(\mathcal{R} + 1)$; (30)

 B^{π} is locally stable if $b < \mathcal{U}^{\pi^2}$ and $0 < \mathcal{U} < \mathcal{U}^{\pi}$ i 1.

B.3 Map $T_{\text{m}}^{(0)}$

The Jacobian matrix at the ... xed point E^{α} , is:

$$
DT_{n}^{(O)}(\mathcal{V}_{n}^{n};\mathcal{V}_{n}^{n}) = \begin{array}{cc} 0 & 1 \\ 0 & 1 \, \mu^{(O)} + \frac{\otimes \mathcal{V}_{n}^{n/2}}{b_{i} \, \mathcal{V}_{n}^{n}} \end{array} \tag{31}
$$

so the eigenvalues are always real, $z_1 = 0$, $z_2 = 1$ _j $\Phi + \frac{\Phi \mathcal{H}^{n2}}{b_1 |\mathcal{H}^{n}}$, and E^n is stable if $b > \mathcal{H}^n(1 + \mathcal{H}^n)$.

 $\bar{\mathbf{u}}$

At B^{α} we have

$$
DT^{(O)}(\frac{b}{1+\frac{1}{4^n}};\frac{b}{1+\frac{1}{4^n}})=\frac{0}{\frac{\Phi b_i \cdot \frac{1}{4^n}(1+\frac{1}{4^n})}{\frac{1}{4^n}(1+\frac{1}{4^n})}} \quad 1 \quad \text{if} \quad \frac{1}{\Phi + \frac{\Phi b}{\frac{1}{4^n}(1+\frac{1}{4^n})}} \quad . \tag{32}
$$

so P(1) > 0 for $b < \mathcal{V}^n(1 + \mathcal{V}^n)$, thus con... rming that at $b = \mathcal{V}^n(1 + \mathcal{V}^n)$ the two ... xed points exchange their stability, and the conditions, i.e. $P(i, 1) > 0$ and 1 i Det > 0 are always satis...ed provided that $\mathcal{U}^{\alpha} > 1$. If $b > \mathcal{U}^{\alpha}(1 + \mathcal{U}^{\alpha})$ then the ...xed point B^{α} is a saddle-point , with $i_1 < z_1 < 0$ and $z_2 > 1$, a straightforward consequence of the inequalities $P(i_1) > 0$, $P(1) < 0$ and $P(0) > 0$. The local stability properties of the two …xed points, for if $\mathcal{U}^{\alpha} > 1$, can be summarized as follows: for $b > \frac{1}{4}$ ($1 + \frac{1}{4}$), E^{π} is stable and B^{π} is unstable; for \mathcal{U}^{α} < b < \mathcal{U}^{α} (1 + \mathcal{U}^{α}), E^{α} is unstable and B^{α} is stable.

B.4 Map T

The Jacobian matrix of the map (15) evaluated at the unique ... xed point E is

$$
DT_{\frac{1}{2}} \frac{1}{1 + \frac{1}{2}}; \frac{1}{1 + \frac{1}{2}} = \frac{0}{i \frac{\frac{1}{2}(1 + \frac{1}{2}) + 1}{b(1 + \frac{1}{2}) + 1}} \frac{1}{1 + \frac{\frac{1}{2}(1 + \frac{1}{2}) + 1}{b(1 + \frac{1}{2}) + 1}}.
$$

the characteristic equation (22) has coe¢cients T r = 1 _i $\Phi + \frac{\Phi}{b(1_{i-1})i}$ and Det = $\frac{\Phi}{b(1_{i-1})i}$. The condition $P(1) > 0$ is always satis...ed, hence the stability conditions reduce to:

$$
2 \, \mathbf{i} \, \mathbf{P} + \frac{2^{\circledast}}{\mathbf{b} \, (1 \, \mathbf{i} \, \mathbf{j} \, \mathbf{i} \, \mathbf{j} \, \mathbf{i}} > 0 \qquad \text{and} \qquad \frac{\mathbf{P} \, \mathbf{i} \, \mathbf{b} \, (1 \, \mathbf{i} \, \mathbf{j} \, \mathbf{j} \, \mathbf{i} \, \mathbf{j}}{\mathbf{b} \, (1 \, \mathbf{i} \, \mathbf{j} \, \mathbf{j} \, \mathbf{i} \, \mathbf{j}} \, \mathbf{i}} < 0 \, : \tag{33}
$$

which are both satis…ed in the set (see …g.7)

$$
\Omega_{\hat{s}}\,=\,\frac{\gamma_2}{(b;\,{}^\circledast)\,2\,\,\Omega j}\,\frac{\mu}{b\,<\,\frac{1}{1\,\,{}_{i\quad,\circ}}}\,\,\text{and}\,\,b\,<\,b_{\hat{f}}(\textcolor{red}{}^\circledast)\,\,\frac{\P\quad\mu}{0\,\,{}_{i\quad,\circ}}\,\,\frac{1}{1\,\,{}_{i\quad,\circ}}\,\,\text{and}\,\,b\,>\,b_{\hat{f}_1}(\textcolor{red}{}^\circledast)\,\,\frac{\P\beta_4}{1\,\,{}_{i\quad,\circ}}\,.
$$

In particular, the equation

$$
\mathsf{b} = \mathsf{b}_{\mathsf{h}}(\mathsf{B}) = \frac{\mathsf{B} + 1}{1 \mathsf{i} \quad \text{(34)}
$$

gives a bifurcation curve at which a subcritical Neimark-Hopf bifurcation occurs.

[INSERT FIG. 7]

C Proof of Proposition 2

Proposition 2 is a straightforward consequence of the following basic properties of linear two-dimensional discrete dynamical systems (see e.g. Lorenz (1993) p. 255)

- ² if the eigenvalues z_1 and z_2 of the Jacobian matrix computed at the …xed point E^{π} are complex conjugate with modulus $jz_1j = jz_2j = \frac{p}{\text{Det}} < 1$, where Det is the Jacobian determinant, then the convergence to the …xed point is oscillatory and the distance $k(x_t; y_t)$ \in \mathbb{F}^n k reduces at a rate proportional to $\sum_{\text{Det}}^{\text{t}}$
- ² if the eigenvalues are real and both inside the unit circle, say $0 < jz_1j < jz_2j < 1$, then the distance $k(x_t; y_t)$ \in ^Ek reduces at a rate proportional to jz_2j^t , and if z_2 is positive then the convergence is monotone in the long run, because the dominant eigenvalue, i.e. the eigenvalue with largest modulus, determines the qualitative behavior of the linear system as t! 1.

Of course, the …rst case occurs if the discriminant $\Delta = Tr^2$; 4Det < 0, and the second if the opposite (weak) inequality holds. In our case, let $B = \frac{\otimes \mu^{\alpha}}{b_1 \mu^{\alpha}}$. Then $Tr^{(1)} = Tr^{(F)}(\mathcal{B}) = Tr^{(O)} = 1$ _i $\mathcal{B} + B\mathcal{H}^{\alpha}$, Det^(F) = B, Det^(I) = B \mathcal{H}^{α} , and Det^(O) = 0. Since $b > \frac{1}{4}$ ^{α} and ∞ 2 (0; 1), then for all policies considered we have T r^(P) > 0. Hence, in the case of real eigenvalues the dominant eigenvalue is positive, given by $z_2^{(P)}=0.5$ \propto T r^(P) + $\frac{P_{\overline{\Delta^{(P)}}}}{\Delta^{(P)}}$ $>$ 0. This means that whenever $\Delta^{(P)} > 0$ we have monotone convergence in the long run. But from the above equalities it follows that

 $\Delta^{(1)} = \Delta^{(F)}$ j $4(\mathcal{V}^{\alpha}$ j $1)B$; $\Delta^{(1)} = \Delta^{(O)}$ j $4B\mathcal{V}^{\alpha}$; $\Delta^{(F)} = \Delta^{(O)}$ j $4B$

and the part (i) of Proposition 2 follows from the fact that B & 0 as \mathcal{R} & 0 :

The binary relations of part (ii) can be easily obtained from the previous equalities, recalling that when convergence is monotone the speed of convergence is given by 0:5 $Tr^{(P)} + \frac{P_{\overline{\Delta^{(P)}}}}{\Delta^{(P)}}$; and when it is oscillatory by $\frac{\mathsf{P}_{\overline{Det}(\mathsf{P})}}{\mathsf{Det}(\mathsf{P})}$. For example, to see that F $\hat{\mathsf{A}}_s$ O, notice that 0:5 $Tr^{(0)} + \frac{D_{\overline{\Delta^{(0)}}}}{\Delta^{(0)}} = Tr^{(0)} = 1$; $\Phi + B\mu^{\pi}$, while $\frac{D_{\overline{\Delta^{(F)}}}}{Det^{(F)}} = \frac{D_{\overline{\Delta^{(F)}}}}{B.4}$

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FIGURE CAPTIONS

Fig.1. A superposition, in the parameter space Ω , of the regions of local stability of the target equilibrium $Eⁿ$ under the three di¤erent policies. The region with left boundary OD is the stability region of E^{α} under I, the one bounded by ABC refers to policy F and the one with left boundary ED refers to policy O. The …gure is obtained with $\mathcal{U}^{\alpha} = 1:5$.

Fig.2. Numerical simulations of the model under policy I with $\mathcal{U}^* = 1.5$; $\mathcal{Q} = 0.4$, $\zeta = 0.5$, $b = 2.75$, and $d = 0.03$, i.e., just after the subcritical Neimark-Hopf bifurcation at which the target in‡ation …xed point E[¤] becomes stable, occurring at $b_h^{(1)}(0:4) = 2:4$. (a) The basin of attraction of E^{α} is represented by the gray region, whereas the white region represents the basin of the "cyclic" attractor $A(2)$ (only partially visible in the ...qure). (b) two sequences of expected in‡ation rates are represented versus time, one generated by an initial condition taken in the grey region of …g. (a) and the other one generated by an initial condition taken in the white region.

Fig.3. Numerical simulations of the model under policy I with $\mathcal{U}^* = 1:5$; $\mathcal{Q} = 0:6$, $\mathbf{b} = 3$, $z = 0.03$, and four di¤erent values of, decreasing from (a) to (d). The gray region represents the basin of the target equilibrium E^{α} . In (a) $\beta = 0.5$, and the basin is entirely included in the region (Ia). In (b) $\epsilon = 0.42$, after the contact between the basin boundary and the line s. In (c) $\epsilon = 0.4195$ at the contact between the basin boundary and the line $x = b_i^2$. In (d) $\epsilon = 0.419$, after the contact between the basin boundary and the line $x = b_i^2$, the basin of E^{α} covers the whole plane, i.e. E^{α} is globally stable.

Fig.4. Numerical simulations of the model with policy F with parameters $V^* = 1.5$; $=$ 0:5, $b = 2:5$, $a = 0:03$, and four di¤erent values of ®, decreasing from (a) to (d), such that the two stable equilibria E^{π} and E coexist. The dark-gray region represents the basin B (E^{π}) of the target equilibrium E^{π} , the light-gray region represents the basin B (E) of the higher in‡ation equilibrium E.

Fig.5. Numerical simulations of the model with policy O with parameters $V^* = 1.5$; \sqrt{s} 0:6, $b = 3:9$, $a = 0:03$, and two di¤erent values of ®, such that the two stable equilibria E^{π} and E coexist. The dark-gray region represents the basin B (E^{π}) of the target equilibrium E^{α} , the light-gray region represents the basin B (E) of the higher in‡ation equilibrium E. (a) For $\mathcal{D} = 0.55$ a chaotic attractor also exists around E, whose basin is represented by the white region. The basin B (E) is formed by two disjoint portions. (b) For $\mathcal{D} = 0.53$ the chaotic attractor no longer exists.

Fig. 6. Numerical computations of the extension of the basin of the target equilibrium E^{r}. All the …gures are obtained with $\frac{1}{4}$ = 1:5, $\frac{2}{1}$ = 0:03 and r = 0:2 (…gures on the left) or $\mathcal{D}=0.6$ (...gures on the right). The di¤erent colors represents di¤erent values of the fraction R of initial conditions which generate trajectories converging to the target equilibrium E^{α} , according to the legend in the …gure.

Fig.7. Stability regions for the …xed point E¸ of the map T¸. The grey-shaded area represents the regions of local stability of E_{\perp} in the parameters space (b; ®).

Fig. 1

Fig. 2

Fig. 3

Fig. 4

Fig. 5

