



Long swings in exchange rates: a stochastic control approach

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Abstract

A regime-switching model to describe the exchange rate dynamics is derived as solution to a stochastic control problem. We assume exchange rates evolve according to some macroeconomic variables (*fundamental*) whose dynamics could be described by a Brownian motion with a state-dependent drift. The local Monetary Authority is assumed to intervene influencing the evolution of the *fundamental*, causing the exchange rate to switch from a depreciating to an appreciating regime (and vice versa). We assume the behaviour of the Monetary Authority can be modeled using an optimal control framework where the state variable is represented by the *fundamental*. The solution of the model allows the determination of an endogenous tolerance band within which the exchange rate freely fluctuates.

Keywords: long swings; stochastic control; monetary authorities' intervention; time-consistent policy

1. Introduction

Modeling exchange rate dynamics has been a major goal of scholars and practitioners. Two main approaches have been used in the recent literature to study exchange rate dynamics: the Markov Switching Model (MSM) and the Random Walk Hypothesis (RWH).

The power of the MSM for exchange rate modeling was first demonstrated by Engel and Hamilton (1990) and Engel (1994), who were able to identify the presence of long swings in exchange rates. Using an econometric approach, they decompose a non-stationary time series into a sequence of stochastic segmented time trends with two regimes (states). The mean of the exchange rate changes may vary across regimes allowing the identification of long swings.

Evidence of long swings behavior in exchange rates have also been reported in Evans and Lewis (1995), Marsh (2000), Bollen et al. (2000), Dewachter (2001) and Clarida et al. (2003),

showing that MSM is able to describe the exchange rate dynamics, but it is unstable over time and not suitable for forecasting.

The RWH was first introduced by Meese and Rogoff (1983) evidencing how the random walk forecasts outperform the results obtained using structural exchange rate models. More recent studies by Meese and Rose (1991), MacDonald and Taylor (1994), Chinn and Meese (1995), MacDonald (1999) and Cheung et al. (2005) and Cheung and Erlandsson (2005) have also reported the inadequacy of traditional exchange rate models, structural and time-series models, to provide accurate forecasts of exchange rates. However, the RWH remains unsatisfactory, from an economic point of view, as it ignores the impact of the *fundamental*.

We believe that 'long swings in the exchange rate pose important challenges for existing theory' (Engel and Hamilton, 1990, p. 689), hence the presence of a long-swinging behavior has important implications for modeling exchange rate dynamics and forces the investigation of what factors may lead to such switching dynamics. Kaminsky (1993), for example, shows that a move from a recessive to an expansionary monetary policy increases the exchange rate depreciation resulting in long swings.

Evans and Lewis (1995) assume that exchange rate switches between appreciating and depreciating regimes are incorporated into rational traders' forecasts of the future exchange rate. They use the monetary model and assume that the *fundamental* switches between a depreciating regime and an appreciating one. They also introduce jumps occurring every time a switch of the *fundamental* occurs; hence, exchange rate dynamics are described by a sequence of mixed jump processes.

The aim of this paper is to capture the swinging behavior of the exchange rate and to provide an economic explanation for it. Exchange rate is too important to be left to the market and hence Monetary Authorities (MMAA) have frequently, and at times, forcefully, intervened using monetary policy instruments to influence the path of their respective currencies. In this context, MMAA's interventions are required to keep exchange rates in line with their long-run *fundamental* value. The fact that most Central Banks attempt to manage their exchange rates stresses an important question, i.e. 'what role can they play in achieving the desired exchange rate goal?'

We assume that domestic MMAA may control the evolution of the *fundamental*; hence, exchange rate dynamics could be better described by a *managed floating regime*. This assumption allows the use of a stochastic optimal control approach to model MMAA's interventions.

Given that in the long run the *fundamental* matters (MacDonald, 1999; Mark, 1995), its dynamics could be described by a stochastic differential equation. Following Krugman (1991, 1992), we assume that the exchange rate is a nonlinear function of the *fundamental* and MMAA want to keep it not too far from its long-run equilibrium level. This may result in changes of monetary policies, which may cause a change of the instantaneous drift of the *fundamental* bringing it closer to its equilibrium level and, at the same time, producing a switch in regime: from a depreciating one (*regime 1*) to an appreciating one (*regime 2*), and vice versa. The switching points are optimally determined as solution of the stochastic control problem. We also contribute to the re-examination of the monetary model in a time-varying coefficients context, showing that regime-switching properties provide substantial explanatory power.

A first validation of the proposed model is obtained using the monthly D-Mark/US-dollar exchange rate over the period 1973–2005. Evidence of regime switches for D-Mark/US-dollar is provided.

The paper is structured as follows: Section 2 presents the stochastic control approach and its solution; Section 3 reports the estimation of the model based on the Method of Simulated Moments and the statistics for the estimated parameters; Section 4 provides some concluding remarks.

2. The theoretical model

We assume the exchange rate is a nonlinear function of the *fundamental* whose dynamics are determined by a set of macroeconomic variables and policy instruments. MMAA set a long-run equilibrium level for the *fundamental*. They monitor the dynamics of the current *fundamental* in order to avoid its ‘undesired’ changes.

In this context, we define an optimal control model that describes MMAA’s behavior. The control variable is represented by the drift of the process used to model the *fundamental*’s dynamics.

2.1. The exchange rate

According to the standard flexible price monetary model, the spot exchange rate at time t is described by

$$s_t dt = f_t dt + \lambda E_t[ds_t], \quad \lambda > 0, \quad (1)$$

where s_t is the logarithm of the exchange rate defined as unit of domestic currency per unit of the reference currency; f_t denotes the logarithm of a set of macroeconomic variables (*fundamental*) driven by a Brownian motion with constant instantaneous variance, σ^2 ; λ can be interpreted as the semi-elasticity of domestic money demand with respect to the interest rate and $E_t[ds_t]$ measures the expected change of the exchange rate at time t . Both s_t and f_t are assumed to be observable at time t (Svensson, 1992).

The *fundamental*, f_t , can be decomposed in an endogenous component, sensitive to MMAA’s intervention, and an exogenous one. The process for the *fundamental*, f_t , is described by the stochastic differential equation (SDE)

$$df_t = \theta(f_t) dt + \sigma dB_t, \quad (2)$$

where $\theta(f_t)$ is the instantaneous drift, $\sigma^2 > 0$ is the instantaneous variance and B_t is a standard Brownian motion.

The theoretical *fundamental*, which represents the long-run equilibrium level, depends on a set of macroeconomic variables that affect the exchange rate. We assume its changes can be described by

$$d\tilde{f}_t = \beta_t dt, \quad \beta_t : \mathfrak{R} \rightarrow \mathfrak{R}. \quad (3)$$

We suppose that MMAA intervene in order to maintain the *fundamental*, f_t , broadly in line with the theoretical value, \tilde{f}_t . Doing so, we may assume MMAA’s aim is to minimize the difference $x_t = f_t - \tilde{f}_t$.

Interventions are justified by Mussa (1981): ‘A valid case for interventions lies on the ground that the authorities may have better knowledge of their own future policy intentions than private market participants’. MMAA can use standard monetary policy tools to modify the *fundamental's* dynamics. Several approaches describe the effects of these interventions on the dynamics of the exchange rates (among others, Miller and Zhang, 1996; Mundaca, 2001).

In this work we assume that MMAA intervene to modify the rate of instantaneous change of the *fundamental*, $\theta(f_t)$, representing the control variable.

Formally, the dynamics of x_t on $(\Omega, \mathfrak{F}, P^{\bar{x}})$ is given by

$$dx_t = df_t - d\tilde{f}_t = [\theta(f_t) - \beta_t] dt + \sigma dB_t, \quad (4)$$

where $\theta(f_t)$ and β_t are bounded functions, and $P^{\bar{x}}$ defines the probability measure of a Brownian motion with initial position $\bar{x} = x_0 \in \mathfrak{R}$. Applying stochastic calculus,¹ we have

$$dx_t = \theta(f_t) dt + \sigma d\hat{B}_t \quad \text{with } d\hat{B}_t = dB_t - \frac{\beta_t}{\sigma} dt, \quad (5)$$

where \hat{B}_t is defined on $(\Omega, \mathfrak{F}, \hat{P})$. The equivalent probability measure \hat{P} is related to P by

$$\frac{d\hat{P}}{dP} \Big|_{\mathfrak{F}_t} = \exp \left[\int_0^t \phi_s dB_s - \frac{1}{2} \int_0^t \phi_s^2 ds \right]. \quad (6)$$

According to (5) the process, x_t , has the same drift, $\theta(f_t)$, of the *fundamental*, f_t , and the Monetary Authority's goal is realized minimizing the expected discounted quadratic deviation of x_t from the origin

$$J(\bar{x}, \theta) = \min_{\theta \in \Theta} \hat{E}^{\bar{x}} \left\{ e^{-\gamma t} \int_0^{\infty} x_t^2 dt \right\} \quad \text{for } \bar{x} = x_0. \quad (7)$$

The minimization is performed setting the optimal drift $\theta^*(f_t) \in \Theta$, with $\Theta =$ set of Borel measurable functions; $\bar{x} = x_0 = f_0 - \tilde{f}_0$ is the initial value of x_t ; γ is the discounting factor ($\gamma > 0$); $\hat{E}^{\bar{x}}$ is the expected value corresponding to a Brownian motion with initial position $\bar{x} \in \mathfrak{R}$.

In terms of optimal stochastic control the problem becomes

$$\begin{aligned} J(\bar{x}, \theta^*) &= \min_{\theta \in \Theta} J(\bar{x}, \theta), \quad \forall \bar{x} \in \mathfrak{R}, \\ \text{s.t. } dx_t &= \theta(f_t) dt + \sigma d\hat{B}_t. \end{aligned} \quad (8)$$

In (5), we assume $\theta(f_t)$ absolutely continuous with a rate of change bounded between θ_1 and θ_2 : $\theta_1 < \theta(f_t) < \theta_2$; $[\theta_1, \theta_2]$ represent the extreme sustainable measures to curb the dynamics of the *fundamental* and the drift $\theta(f_t)$ is equal to θ_1 , $\theta_1 < 0$ when contractionary monetary policies are introduced and it is equal to θ_2 , $\theta_2 > 0$ when expansionary monetary policies are adopted.

¹According to Girsanov's Theorem it is possible to estimate the probability measure of the new Brownian motion, \hat{B}_t .

2.2. The solution of the stochastic control problem

The formal Hamilton–Jacobi–Bellman equation to solve the control problem (8) is

$$\gamma J = x^2 + \frac{\sigma^2}{2} J_{xx} + \inf_{\theta_1 < \theta(f_t) < \theta_2} \{ \theta(f_t) J_x \}, \tag{9}$$

where subscripts identify derivatives.

We expect a solution based on the sign of J_x and we look for two numbers $\delta_i(\theta_j, \gamma)$, for $i = 1, 2$; $j = 0, 1, 2$, with $\delta_1(\cdot) > 0$ and $\delta_2(\cdot) < 0$, identifying the switching points of the process x_t . More precisely, $\delta_1(\cdot)$ and $\delta_2(\cdot)$ represent the upper and lower threshold values that define an endogenous time-varying ‘tolerance band’ and a time-consistent policy. We assume that if x_t exceeds the threshold values, $\delta_i(\cdot)$ ($i = 1, 2$), MMAA’s interventions occur:

- if $x_t \geq \delta_1(\cdot) > 0$ restrictive monetary policies are adopted causing the *fundamental* to switch from an appreciating regime to a depreciating one;
- if $x_t \geq \delta_2(\cdot) < 0$ expansionary policies are introduced causing the switch from the depreciating regime to the appreciating one;
- if $\delta_2(\cdot) < x_t < \delta_1(\cdot)$ no interventions occur and the *fundamental* follows the dynamics described by equation (2).

According to the initial position of x , $\bar{x} = x_0$, the corresponding formalization of (9) is

$$\begin{aligned} (1) \quad & \gamma J = x^2 + \frac{\sigma^2}{2} J_{xx} + \theta_2 J_x, \quad J_x \leq 0, \quad \bar{x} \leq \delta_2(\cdot), \\ (2) \quad & \gamma J = x^2 + \frac{\sigma^2}{2} J_{xx} + \theta(f_t) J_x, \quad \delta_2(\cdot) < \bar{x} < \delta_1(\cdot), \\ (3) \quad & \gamma J = x^2 + \frac{\sigma^2}{2} J_{xx} + \theta_1 J_x, \quad J_x \geq 0, \quad \bar{x} \geq \delta_1(\cdot), \end{aligned} \tag{10}$$

where we assume that whenever \bar{x} finds itself within the tolerance band, $]\delta_2(\cdot), \delta_1(\cdot)[$, the drift of the process is $\theta(f_t)$.

We solve system (10) – details are reported in the Appendix – to derive the threshold values, $\delta_1(\cdot)$ and $\delta_2(\cdot)$, and the optimal drift, $\theta^*(f_t)$, for the process described by (5)

$$\theta^*(f_t) = \begin{cases} \theta(f_t) & \delta_2 < \bar{x} < \delta_1 \\ \theta_1 & \bar{x} \geq \delta_1 \\ \theta_2 & \bar{x} \leq \delta_2 \end{cases} \quad \bar{x} = x_0 \tag{11}$$

Solution (11) says that MMAA push as hard as possible to reduce the difference $x_t = f_t - \tilde{f}_t$. This means to move x_t to the right of the critical point $\delta_2(\cdot)$ when $\bar{x} \leq \delta_2$ and, vice versa, to the left whenever $\bar{x} \geq \delta_1(\cdot)$. Given the definition of x_t (4), the *fundamental* dynamics become

$$\theta^*(f_t) = \begin{cases} \theta(f_t), & \tilde{f}_t + \delta_2(\cdot) < f_0 < \delta_1(\cdot) + \tilde{f}_t, \\ \theta_1, & f_0 \geq \delta_1(\cdot) + \tilde{f}_t, \\ \theta_2, & f_0 \leq \delta_2(\cdot) + \tilde{f}_t, \end{cases} \tag{12}$$

where f_0 is the initial point of the current *fundamental*.

Expression (12) says that MMAA’s aim is to move f_t to the right (left) of the critical point $\delta_2(\cdot) + \tilde{f}_t$ ($\delta_1(\cdot) + \tilde{f}_t$) when $f_t \leq \delta_2(\cdot) + \tilde{f}_t$ ($f_t \geq \delta_1(\cdot) + \tilde{f}_t$).

We assume that whenever f_t exceeds the tolerance band ($f_t \notin [\delta_2(\cdot) + \tilde{f}_t, \delta_1(\cdot) + \tilde{f}_t]$), some kind of monetary policy interventions occur aimed at bringing f_t back into the tolerance band, while no action is taken if f_t moves within the tolerance band. The intervention generates an inversion in the trend of the *fundamental* and causes a switch in regime (state), bringing the process again within the tolerance band according to the SDE (2).

2.3. The exchange rate dynamics

In order to express equation (1) in state-space rather than time-series representation, we consider its solution as a continuous twice-differentiable function of the *fundamental*

$$s_t = h(f_t). \tag{13}$$

Applying Ito's lemma we obtain the dynamics of s , which is described by²

$$ds = h'(f) df + \frac{1}{2}h''(f) [df]^2 = h'(f)[\theta(f) dt + \sigma dB] + \frac{1}{2}h''(f)\sigma^2 dt. \tag{14}$$

Taking the conditional expectation of (14) and using the law of motion (1) we obtain

$$E_t ds = h' E_t \theta(f) dt + \frac{1}{2}h'' \sigma^2 dt = \frac{1}{\lambda}(s - f) dt. \tag{15}$$

Solving (15) we obtain an expression for the exchange rate related to the *fundamental* described by (2).³ Using Krugman's approach (1991, 1992), we obtain the general closed form solution of (15)

$$s = h(f) = \begin{cases} f + \lambda\theta_1 + G_1 e^{v_1 f} + I_1 e^{o_1 f}, & f_0 \geq \delta_1(\cdot), \\ f + \lambda\theta_0 + G_0 e^{v_0 f} + I_0 e^{o_0 f}, & \delta_2(\cdot) < f_0 < \delta_1(\cdot), \\ f + \lambda\theta_2 + G_2 e^{v_2 f} + I_2 e^{o_2 f}, & f_0 \leq \delta_2(\cdot), \end{cases} \tag{16}$$

where we assume that within the tolerance band the drift is constant and equal to $\theta_0 = \theta(f_t)$, and

$$\begin{aligned} v_i &= \sigma^{-2} \left(-\theta_i + \sqrt{\theta_i^2 + 2\frac{\sigma^2}{\lambda}} \right) > 0 \\ o_i &= \sigma^{-2} \left(-\theta_i - \sqrt{\theta_i^2 + 2\frac{\sigma^2}{\lambda}} \right) < 0 \end{aligned} \quad i = 0, 1, 2 \tag{17}$$

In (16) we need to determine G_i ($i = 0, 1, 2$) and I_i ($i = 0, 1, 2$) according to the following boundary conditions

$$\begin{aligned} (1) \quad & G_0 = 0, I_0 = 0, \\ (2) \quad & G_1 = 0, I_2 = 0, \\ (3) \quad & \begin{cases} \lambda\theta_1 + I_1 e^{o_1 \delta_1} = \lambda\theta_0, \\ \lambda\theta_2 + I_2 e^{o_2 \delta_2} = \lambda\theta_0, \end{cases} \end{aligned} \tag{18}$$

where

²To simplify notations we omit, where possible, time dependence.

³To simplify notations we assume, without loss of generality, that the initial level for the theoretical *fundamental* is $\tilde{f}_0 = 0$.

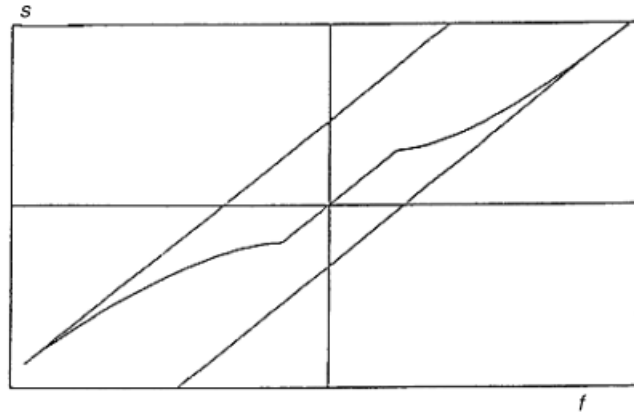


Fig. 1. Exchange rate dynamics within the tolerance band.

- condition (1) implies that inside the tolerance band $[\delta_2(\cdot), \delta_1(\cdot)]$ MMAA do not adopt any policy measure to modify the dynamics of the *fundamental*, hence the correct solution is defined by the linear component of (15);
- condition (2) derives from the asymptotic behaviour of s :

$$\begin{aligned} \lim_{f \rightarrow +\infty} h(f) &= f + \lambda\theta_1, \\ \lim_{f \rightarrow -\infty} h(f) &= f + \lambda\theta_2, \end{aligned} \tag{19}$$

which varies within the band defined by $f + \lambda\theta_1$ and $f + \lambda\theta_2$;

- condition (3) avoids the occurrence of discrete jumps at $f = \delta_i$ ($i = 1, 2$).

The final solution, therefore, is defined by

$$\begin{aligned} s &= f + \lambda\theta_0, & \delta_2 < f < \delta_1, \\ s &= f + \lambda\theta_1 - \lambda(\theta_1 - \theta_0)e^{\rho_1(f - \delta_1)}, & f \geq \delta_1, \\ s &= f + \lambda\theta_2 - \lambda(\theta_2 - \theta_0)e^{\rho_2(f - \delta_2)}, & f \leq \delta_2. \end{aligned} \tag{20}$$

Figure 1 describes the dynamics of the exchange rate, s , solution of (15).

3. Model's validation

In this section we provide a validation of the model described in Section 2. We use monthly data for the D-Mark/US-dollar exchange rate over the period January 1973 (the beginning of the general floating exchange rate system) to January 2005.

To estimate the parameters characterizing the *fundamental* dynamics we use the simulated moments estimation (SME). The estimated parameters have been statistically tested.

We use the following specification for the monetary model:

$$s_t = f_t + \lambda[E_t(s_{t+1} - s_t)], \tag{21}$$

Table 1
OLS results for equation (24)

g	t	σ_g	R^2
0.8	2.28	0.054	0.67

where λ is the semi-elasticity of domestic money demand; s_t is the logarithm of the spot exchange rate (in terms of home currency, D-Mark, per unit of foreign currency, US-dollar); E_t is the conditional expectations operator defined by $E_t(s_{t+1}) \equiv E_t(s_{t+1}|\mathfrak{S}_t)$; f_t defines the *fundamental*.

We assume the observability of the *fundamental* (Svensson, 1992). In addition, given the Uncovered Interest Parity holds,⁴ f_t is obtained from (21) as the difference between the spot exchange rate, s_t , and the one-month interest rate differential ($r_t - r_t^*$):

$$\hat{f}_t = s_t - \hat{\lambda}(r_t - r_t^*). \quad (22)$$

According to Huang (1981), a realistic estimation of the *fundamental* is obtained assuming $\hat{\lambda} = 1.5$.

In this model, MMAA's interventions are introduced when the difference $x_t = f_t - \tilde{f}_t$ exceeds the tolerance band $]\delta_2(\cdot), \delta_1(\cdot)[$. The threshold values, $\delta_1(\cdot)$ and $\delta_2(\cdot)$, and the strength of the control, $\theta(f_t)$, in (5) are determined as the solution of the stochastic control problem described by (8).

We estimate the theoretical level \tilde{f}_t using the \hat{f}_t described by (22) and assuming the term β_t in (3) is defined by

$$\beta_t = g(r_t - r_t^*). \quad (23)$$

Standard OLS technique on the difference of \hat{f}_t allows the estimation of g :

$$\Delta \hat{f}_t = g(\Delta r_t - \Delta r_t^*) + \varepsilon_t. \quad (24)$$

The results of the estimation of (24) are reported in Table 1.

3.1. Estimation of the fundamental

The process for the *fundamental*, f_t , is generated by the difference equation

$$f_t^\xi = h(f_{t-1}; \varepsilon_t; \xi_0), \quad (25)$$

where h is the transition function; $\xi_0 = (\theta_1, \theta_2, \theta^+, \theta^-, \sigma, \gamma)$ is the unknown q -dimensional vector and $\{\varepsilon_t\}$ is an i.i.d. sequence of $N(0, 1)$ random variables, defined on the probability space $(\Omega, \mathfrak{S}, P)$. To calibrate the model and estimate $\theta^*(f_t)$ according to (12), we make some realistic assumptions about $\theta(f_t)$. Precisely, if the process, f_t , moves within the tolerance band $\tilde{f}_t + \delta_2 < f_t < \delta_1 + \tilde{f}_t$, after the adoption of expansionary (contractionary) policies, the drift of the process $\theta(f_t)$ assumes a constant positive (negative) value: θ^+ (θ^-).

From (25) a set of simulated equilibrium states, $\{f_t^\xi\}_{t=1}^N$, can be generated. For some chosen observation function h , in each period t an observation $Z = h(f_t, f_{t-1}, \dots, f_{t-l-1})$ is made of a

⁴According to the UIP: $E_t(s_{t+1} - s_t) = r_t - r_t^*$.

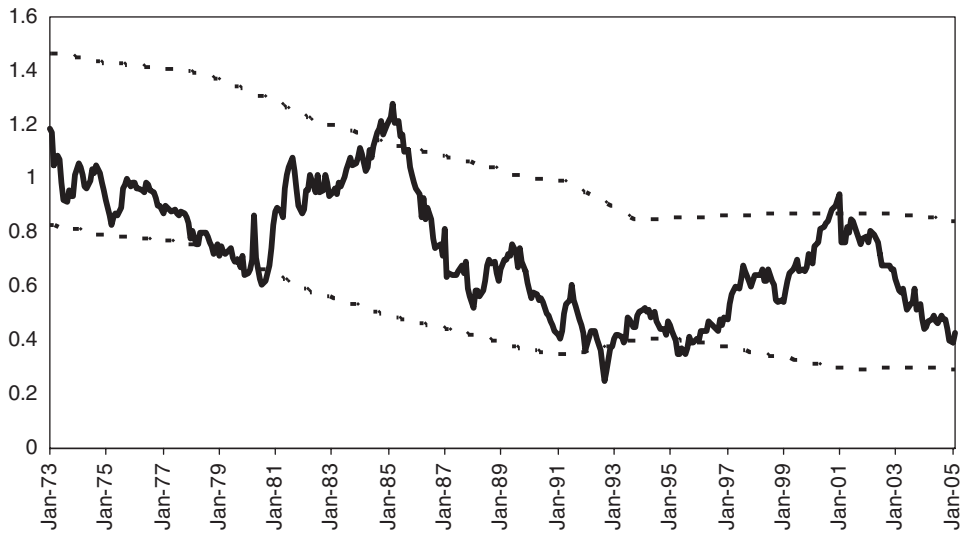


Fig. 2. Fundamental dynamics within and outside the tolerance band.

finite ‘1-history’ of state information. A corresponding observation of $\{f_t^\xi\}$ can be formed for each ‘1-history’ of simulated states. Finally, the SME sets a value of ξ chosen to minimize the distance between the sample first three moments of $\{f_t^\xi\}_{t=1}^N$ and the sample first three moments of $\{\hat{f}_t\}_{t=1}^T$, where T is the number of historical observations (Duffie and Singleton, 1993; Newey, 2001). The same procedure is repeated on the first differences of f , Δf .

For each initial point $\hat{f}_0 = f_0$ and each parameter vector $\xi \in \Xi$, the simulated state process (\hat{f}_t^ξ) is built as

$$\begin{aligned} \hat{f}_0^\xi &= \hat{f}_0, \\ \hat{f}_{t+1}^\xi &= h(\hat{f}_t^\xi, \hat{\varepsilon}_{t+1}, \xi). \end{aligned} \tag{26}$$

Using the SME, we compute the first three moments, μ_1, μ_2, μ_3 of $\{f_t^\xi\}_{t=1}^N$, and the first three moments, μ_1', μ_2', μ_3' of $\{\Delta f_t^\xi\}_{t=1}^N$. Hence we have an exactly identified case. We run 1200 simulations and estimate the vector $\xi_0 = (\theta_1, \theta_2, \theta^+, \theta^-, \sigma, \gamma)$, obtaining

$\hat{\theta}_1$	$\hat{\theta}_2$	$\hat{\theta}^+$	$\hat{\theta}^-$	$\hat{\sigma}$	$\hat{\gamma}$
-0.0376	0.03469	0.003114	-0.00385	0.0922	0.1185

Solving the system of nonlinear equations (16), we obtain the estimated tolerance band, $\delta_i(\theta_j, \gamma),]-0.3052; 0.43196[$. In Fig. 2 the dynamics for the estimated *fundamental* within the tolerance band $]-0.30052 + \tilde{f}_i; 0.43196 + \tilde{f}_i[$ is presented.

These results imply:

- if $f_t - \tilde{f}_i \leq -0.3052$, an intervention by MMAA occurs, i.e. expansionary monetary policies generate a positive drift, $\theta_2 = 0.03469$.

Table 2

Actual and simulated moments: test statistic for the *fundamental*

	Actual	Simulated	% of simulated cases exceeding the reference ones
$\mu_1 (f)$	0.824901	0.806774	48
$\mu_2 (f)$	0.839948	0.731886	26
$\mu_3 (f)$	0.653658	0.720764	71
Skewness (f)	-0.14727	-0.01183	22
Kurtosis (f)	-0.71005	-1.15951	89
$\mu'_1 (\Delta f)$	-0.00054	-0.00058	51
$\mu'_2 (\Delta f)$	0.00288	0.0007483	75
$\mu'_3 (\Delta f)$	-0.0006	-0.0026	79
Skewness (Δf)	-0.12032	-0.00191	6
Kurtosis (Δf)	1.45574	0.295208	4
G		0.0031145	

- if $f_t - \tilde{f}_t \geq 0.43196$, restrictive monetary policies are adopted and the action is reflected in a negative drift for f_t , $\theta_1 = -0.0376$.
- if $-0.3052 < f_t - \tilde{f}_t < 0.43196$, MMAA let the *fundamental* freely float inside the tolerance band. We estimate a positive drift, $\theta_0^+ = 0.003114$, and a negative one, $\theta_0^- = 0.00385$, for the appreciating and depreciating regimes, respectively.

The implementation of SME is based on a number of sample moments exactly equal to the number of parameters to be estimated; hence we could not test the estimated model using the over-identifying restriction. As an alternative, we adopt Smith and Spencer's method (1992) and generate n replications using the estimated parameters. In each replication μ_1, μ_2, μ_3 and μ'_1, μ'_2, μ'_3 were estimated, together with a test for normality. The results are reported in Table 2, where the sets of the first three moments on the levels and on the first differences of the observed and the simulated series are listed. On each of these series the statistics for skewness and kurtosis are reported. The moments for the simulated series were obtained using $n = 100$ replications.

In the last column of Table 2 the percentage of replications in which the simulated value is greater than the observed one, for positive sample statistics, or is less than the observed value, for negative ones, is also reported. On average, 51 out of 100 replications provide a mean change in the *fundamental* lower than the change in the observed series.⁵ All the values in Table 2 support the statistical significance of the estimated parameters. The only non-significant statistic refers to the kurtosis of the first differences, showing that the model does not succeed in describing the observed kurtosis.

⁵A statistic is considered well matched if the proportion of replications in which the simulated value exceeds the historical value falls roughly within the interval (0.05, 0.95).

4. Conclusions

The paper presents a new theoretical approach to describe the long-swinging behavior of exchange rates observed over the last 30 years.

The dynamics of the exchange rates are described as a function of the macroeconomic *fundamental*, f_t , which follows a Brownian motion with state-dependent drift. The process for the dynamics of the *fundamental* is obtained as the solution of a stochastic control problem, which describes the MMAA's aim to keep the value of the *fundamental* as close as possible to a predefined long-run path. The optimal control approach allows one to set an endogenous time-varying tolerance band, outside of which an optimal time-consistent policy takes place.

The approach is realistic because it suggests a more adequate process to describe exchange rate dynamics and provides an accurate analysis of the observed phenomenon with respect to simple diffusion processes or MSMs, which may lack economic explanation. The model takes into account MMAA's role in preventing the exchange rate to be set at wrong levels and shows that a change in monetary policy may lead to the long-swinging behavior of exchange rates. Furthermore, we contribute to the re-examination of the monetary model in a time-varying context and to the definition of possible interactions between the exchange rates and the *fundamental*.

An application of the model is also provided. The estimation of the parameters that define the dynamics of the *fundamental* supports the optimal control approach and the long-swinging behavior of the exchange rates. The model succeeds in capturing all the major switches in regime observed in the D-Mark/US-dollar exchange rate over the period 1973–2005.

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Appendix A

We look for solutions $J = O(x^2)$ as $|x| \uparrow \infty$ of (10) which meet smoothly at $x = \delta_2$ and $x = \delta_1$ to order 1, since the cost of control increases nonlinearly when x_t is far from the origin. As a particular solution of the first equation in (10) we find

$$\frac{x^2}{\gamma} + \frac{\sigma^2}{\gamma^2} + \frac{2\theta_2 x}{\gamma^2} + \frac{2\theta_2^2}{\gamma^3} \quad \bar{x} \leq \delta_2. \quad (\text{A.1})$$

A general solution of (10) is obtained adding to (A.1) a solution of $\gamma J = \frac{\sigma^2}{2} J_{xx} + \theta_2 J_x$ given by

$$C_2 e^{\lambda_2 x} + D_2 e^{\mu_2 x}, \quad (\text{A.2})$$

where

$$\begin{aligned} \lambda_2 &= \sigma^{-2} \left(-\theta_2 + \sqrt{\theta_2^2 + 2\gamma\sigma^2} \right) > 0, \\ \eta_2 &= \sigma^{-2} \left(-\theta_2 - \sqrt{\theta_2^2 + 2\gamma\sigma^2} \right) < 0, \end{aligned} \tag{A.3}$$

and C_2, D_2 are arbitrary constants.

A similar approach is applied for solutions of the other two equations in (10), so the general solution is

$$\gamma J = \begin{cases} \frac{x^2}{\gamma} + \frac{\sigma^2}{\gamma^2} + \frac{2\theta_2 x}{\gamma^2} + \frac{2\theta_2^2}{\gamma^3} + C_2 e^{\lambda_2 x} + D_2 e^{\eta_2 x}, & \bar{x} \leq \delta_2, \\ \frac{x^2}{\gamma} + \frac{\sigma^2}{\gamma^2} + \frac{2\theta_0 x}{\gamma^2} + \frac{2\theta_0^2}{\gamma^3} + C_0 e^{\lambda_0 x} + D_0 e^{\eta_0 x}, & \delta_2 < \bar{x} < \delta_1, \\ \frac{x^2}{\gamma} + \frac{\sigma^2}{\gamma^2} + \frac{2\theta_1 x}{\gamma^2} + \frac{2\theta_1^2}{\gamma^3} + C_1 e^{\lambda_1 x} + D_1 e^{\eta_1 x}, & \bar{x} \leq \delta_1, \end{cases} \tag{A.4}$$

where terms λ_i and η_i for $(i = 0, 1, 2)$ are defined as in (A.3), using the corresponding drifts $\theta_j, j = 0, 1, 2$. The final expression for the threshold values $\delta_1(\theta_j, \gamma)$ and $\delta_2(\theta_j, \gamma)$ is obtained solving (A.4) and defining the six arbitrary constants, $C_i (i = 0, 1, 2)$ and $D_i (i = 0, 1, 2)$.

We are able to obtain the values for δ_1 and δ_2 as $f(\theta_0, \theta_1, \theta_2, \gamma)$ solving a system of nonlinear equations

$$\begin{cases} a_{11} e^{\lambda_0 \delta_2 - \lambda_0 \delta_1} + a_{12} e^{\eta_2 \delta_2 - \eta_0 \delta_1} = b_{11}, \\ a_{21} e^{\lambda_0 \delta_2 - \lambda_0 \delta_1} + a_{22} e^{\eta_2 \delta_2 - \eta_0 \delta_1} = b_{21}, \end{cases} \tag{A.5}$$

Equation (A.5) is obtained substituting the value of the constants $C_0, C_1, C_2, D_0, D_1, D_2$ in expression (A.4).

$C_i (i = 0, 1, 2)$ and $D_i (i = 0, 1, 2)$ are obtained imposing the quadratic growth condition; for $|x| \uparrow \infty, C_2 = 0$ and $D_1 = 0$. The other four arbitrary constants are obtained assuming continuity of $J(x)$ and its first derivative, $J_x(x)$, in $x = \delta_2$ and $x = \delta_1$, so we get the following four equations to determine $C_1, C_0, D_0,$ and D_2 .

$$\begin{cases} \frac{2\theta_0}{\gamma^2} - \frac{2\theta_1}{\gamma^2} + C_0 \lambda_0 e^{\lambda_0 \delta_1} - C_1 \lambda_1 e^{\lambda_1 \delta_1} + D_0 \eta_0 e^{\eta_0 \delta_1} = 0, \\ \frac{2(\theta_0 - \theta_1)}{\gamma^2} \delta_1 + \frac{2\theta_0^2}{\gamma^3} - \frac{2\theta_1^2}{\gamma^3} + C_0 e^{\lambda_0 \delta_1} - C_1 e^{\lambda_1 \delta_1} + D_0 \eta_0 e^{\eta_0 \delta_1} = 0, \\ \frac{2\theta_0}{\gamma^2} - \frac{2\theta_2}{\gamma^2} + C_0 \lambda_0 e^{\lambda_0 \delta_2} - D_2 \eta_2 e^{\eta_2 \delta_2} + D_0 \eta_0 e^{\eta_0 \delta_2} = 0, \\ \frac{2(\theta_0 - \theta_2)}{\gamma^2} \delta_2 + \frac{2\theta_0^2}{\gamma^3} - \frac{2\theta_2^2}{\gamma^3} + C_0 e^{\lambda_0 \delta_2} - D_2 e^{\eta_2 \delta_2} + D_0 e^{\eta_0 \delta_2} = 0. \end{cases} \tag{A.6}$$

Solving system (A.6) and assuming also continuity of the second derivative $J_{xx}(x)$ for $x = \delta_2$ and $x = \delta_1$, we get an expression for C_1 and D_2 as function of C_0 and D_0

$$\begin{cases} C_1 \lambda_1 e^{\lambda_1 \delta_1} = D_0 \eta_0^2 e^{\eta_0 \delta_1} + C_0 \lambda_0^2 e^{\lambda_0 \delta_1}, \\ D_2 \eta_2^2 e^{\eta_2 \delta_2} = D_0 \eta_0^2 e^{\eta_0 \delta_2} + C_0 \lambda_0^2 e^{\lambda_0 \delta_2}. \end{cases} \tag{A.7}$$

Substituting (A.7) in (A.6), after some algebra, we get the final expressions for C_0 , C_1 , D_0 and D_2

$$\begin{aligned} C_0 &= \frac{p_1 \lambda_1 \eta_0 - q_1 (\lambda_1 - \eta_0)}{(\eta_0 - \lambda_0)(\lambda_1 - \lambda_0)} e^{-\lambda_0 \delta_1}, \\ C_1 &= \frac{C_0 \lambda_0^2 e^{(\lambda_0 - \eta_2) \delta_2} + D_0 \eta_0^2 e^{(\eta_0 - \eta_2) \delta_2}}{\eta_2^2}, \\ D_0 &= \frac{p_1 \lambda_1 \lambda_0 - q_1 (\lambda_1 + \lambda_0)}{(\lambda_0 - \eta_0)(\eta_1 - \eta_0)} e^{-\eta_0 \delta_1}, \\ D_2 &= \frac{C_0 \lambda_0^2 e^{(\lambda_0 - \lambda_1) \delta_1} + D_0 \eta_0^2 e^{(\eta_0 - \lambda_1) \delta_1}}{\lambda_1^2}, \end{aligned} \tag{A.8}$$

with

$$\begin{aligned} q_1 &= \frac{2}{\gamma^2} (\theta_0 - \theta_1), \\ q_2 &= \frac{2}{\gamma^2} (\theta_0 - \theta_2), \\ p_1 &= \delta_1 q_1 + \frac{2}{\gamma^3} (\theta_0^2 - \theta_1^2), \\ p_2 &= \delta_2 q_2 + \frac{2}{\gamma^3} (\theta_0^2 - \theta_2^2). \end{aligned} \tag{A.9}$$

Furthermore, it could be verified that non-negative constants C_1 and D_2 guarantee the convexity of function $J(x)$.

Once defined the arbitrary constants C_0 , C_1 , D_0 and D_2 , in order to get the expression for δ_2 and δ_1 as $f(\theta_0, \theta_1, \theta_2, \gamma)$, we impose the values of the solutions for $J(x)$ to match at $x = \delta_2$ and $x = \delta_1$, and we get the following system of nonlinear equations

$$\begin{cases} a_{11} e^{\lambda_0 \delta_2 - \lambda_0 \delta_1} + a_{12} e^{\eta_2 \delta_2 - \eta_0 \delta_1} = b_{11}, \\ a_{21} e^{\lambda_0 \delta_2 - \lambda_0 \delta_1} + a_{22} e^{\eta_2 \delta_2 - \eta_0 \delta_1} = b_{21}, \end{cases} \tag{A.10}$$

with

$$\begin{aligned} a_{11} &= l[q_2(\lambda_0 + \eta_2)], \\ a_{12} &= j[p_2 \eta_2^2 - (\eta_0 + \eta_2)(q_2 - \eta_2 p_2)], \\ a_{21} &= l[-q_1(\lambda_1 + \eta_0) + \lambda_1 \eta_0 p_1], \\ a_{22} &= j[-q_1(\lambda_0 + \eta_2) + \lambda_0 \lambda_1 p_1], \\ b_{11} &= k[q_1 - \lambda_0 p_1], \\ b_{21} &= m[(q_2 - \eta_2 p_2)(\lambda_0 + 2\lambda_1 + \eta_0)], \end{aligned} \tag{A.11}$$

where

$$\begin{aligned}j &= (\eta_0 - \eta_2)(\lambda_0 - \lambda_1), \\k &= (\lambda_1 - \eta_0)(\lambda_0 - \lambda_1), \\l &= (\lambda_1 - \eta_0)(\lambda_0 - \eta_2), \\m &= (\eta_0 - \eta_2)(\lambda_0 - \eta_2),\end{aligned}\tag{A.12}$$

which is solved using Newton's algorithm.