

# **A non-linear multivariate model of the UK real exchange rate**

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## **Abstract**

This paper discusses linearity testing for the UK real exchange rate within a multivariate framework. First we estimate a long-run real exchange rate relationship within a system involving real wages, the unemployment rate and the real price of oil. Then we adopt a logistic transition function for the estimated relationship and show that non-linearities in the discrepancy between the real exchange rate and its implied long-run level affect the short-run real exchange rate equation. We also find that when the real exchange rate is undervalued, unemployment falls as firms respond to an improvement in domestic competitiveness by increasing their demand for labour. At the same time, workers respond to the improvement in domestic competitiveness by demanding and getting higher wages. Further, the effect on unemployment and wages is non-linear.

*Keywords:* Real exchange rate; Traded goods; Smooth Transition Vector Error Correction Model; Quadratic logistic function.

*JEL classification:* C32; C51; C52; F30; F41

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## 1. Introduction

The use of non-linear models in explaining economic phenomena is motivated by the idea that the behaviour of economic variables depends on different states of the world or regimes that prevail at any point in time. Over the last few years, the Smooth Transition Autoregressive (thereafter STAR) methodology has been a popular way of introducing regime-switching behaviour in economic models, where the transition from one regime to the other occurs in a smooth way. STAR models were introduced by Teräsvirta and Anderson (1992) in order to examine non-linearities over the business cycle, whereas their statistical properties were discussed in Granger and Teräsvirta (1993) and Teräsvirta (1994), among others. Extensions of these models to a Vector Autoregressive (VAR) framework were recently discussed in Granger and Swanson (1996), Weise (1999), van Dijk *et al.* (2001) and Rothman *et al.* (2001) among others. However, as Franses and van Dijk (2000, p. 252) point out, "... the analysis of multivariate nonlinear models has been taken up only very recently, and at the time of writing there are no generally accepted ideas on how to construct such multivariate models in the first place".

This paper attempts to construct and estimate a multivariate STAR model for the UK real exchange rate. The starting point of our analysis is an earlier paper by Alogoskoufis (1990) who derived a real exchange rate equation based on the production sector of the economy which distinguishes between traded and non-traded goods. Alogoskoufis (1990) estimated the real exchange rate equation by instrumental variable estimation as part of a system also involving a domestic output and a real wage equation. We extend the work of Alogoskoufis in at least two directions. First, we apply the Johansen (1988) cointegrating technique to estimate jointly the long-run properties and the short-run dynamics of the real exchange rate equation as part of a small system involving a measure of the real exchange rate, real wages, the unemployment rate

and the real price of oil. Second, we look at the possibility that deviations of the real exchange rate equation from its long run level affect the short-run dynamics of the system in a non-linear way using the STAR methodology. STAR models allow for regime-switching behaviour where the transition between regimes occurs in a smooth way. In that respect, STAR models contrast with the Threshold Autoregressive (TAR; see e.g. Tong, 1990) and the Hamilton (1989) Markov regime-switching models, which assume that the transition between regimes occurs abruptly rather than smoothly. On economic grounds, STAR models seem to be more appropriate than TAR or Markov regime-switching models for modelling non-linearities in the adjustment of the real exchange rate. Modelling the real exchange rate as a function of real wages and the real price of oil in the rest of this paper implies that real exchange rate movements are affected by conditions prevailing in the production sector of the economy. In this case, a smooth transition rather than a sharp switch between regimes could be justified in terms of frictions in the product market due to product heterogeneity, government imposed barriers to trade, or labour market inflexibility distorting the rapid adjustment of wages.

When the transmission mechanism is controlled by the deviations of the real exchange rate equation from its long-run equilibrium level, we can differentiate between two regimes. The first one is associated with (time periods of) small deviations of the real exchange rate from its long run level. This implies the existence of an interval band around the equilibrium rate in which there is no tendency for the real exchange rate to move back to its equilibrium quickly. If this is the case, then the impact of the disequilibrium error on the short-run dynamics of the system is expected to be small. The second regime is associated with (time periods of) large deviations of the real exchange rate from its long-run level. In this case, there is a strong tendency for the real exchange rate to move back to its equilibrium quickly. As a result, the impact of the disequilibrium error on the short-run dynamics of the system is expected to be large.

Our estimates suggest the existence of a long-run real exchange rate equation which affects significantly the short-run dynamics of the system both in a linear and a non-linear way. In particular, the short-run real exchange rate adjusts quickly to disequilibrium deviations of the real exchange rate from its optimal level outside an interval band, which is estimated to be rather wide. This is not surprising as our sample period coincides with floating exchange rates being in operation. We also find that when the real exchange rate is above its long-run equilibrium level (i.e. it is undervalued), short-run unemployment falls as firms respond to an improvement in domestic competitiveness by increasing their demand for labour. Further, there is a strong response of short-run unemployment to the disequilibrium error outside an interval band, which is estimated to be rather narrow. To the extent that the real exchange rate equation reflects monetary and more generally economic policy considerations our results imply that unemployment can be targeted by economic policy. Furthermore, if economic authorities want to avoid large swings in unemployment then they should be prepared to intervene in exchange markets with the aim of keeping real exchange rate movements within a narrow interval band. Our results also suggest that when the real exchange rate is undervalued, workers respond to an improvement in domestic competitiveness by demanding and getting higher wages. Again, this effect is non-linear. Therefore, our findings recognise an important role for the real exchange rate in affecting labour market conditions in the UK.

The structure of the paper is as follows. The next section discusses briefly the theory of linearity testing within a STVECM framework. Section 3 of the paper discusses the econometric specification of a real exchange rate model, whereas Section 4 estimates the linear and non-linear versions of the model. Section 5 presents a discussion of our findings and section 6 provides some concluding remarks.

## 2. Specification of STVECM models

Following Rothman *et al.* (2001), we write a  $k$ -dimensional Smooth Transition Vector Error Correction Model (STVECM) as:

$$\begin{aligned} \Delta y_t = & \left( \mu_1 + \alpha_1 z_{t-1} + \sum_{j=1}^{p-1} \Phi_{1,j} \Delta y_{t-j} + \sum_{j=0}^{p-1} \Phi_{2,j} \Delta x_{t-j} \right) (1 - G(s_t)) \\ & + \left( \mu_2 + \alpha_2 z_{t-1} + \sum_{j=1}^{p-1} \Phi_{3,j} \Delta y_{t-j} + \sum_{j=0}^{p-1} \Phi_{4,j} \Delta x_{t-j} \right) G(s_t) + \varepsilon_t, \end{aligned} \quad (1)$$

where  $y_t$  is a  $(k \times 1)$  vector of  $I(1)$  endogenous variables,  $x_t$  is an  $(m \times 1)$  vector of  $I(1)$  exogenous variables,  $\varepsilon_t \sim iid(0, \Sigma)$ ,  $\alpha_i$ ,  $i = 1, 2$ , are  $(k \times r)$  matrices, and  $z_t = \beta' [y_t', x_t']'$  for some  $(q \times r)$  matrix  $\beta$  denote the error correction terms, with  $q = k + m$ .  $\Phi_{1,j}$  and  $\Phi_{3,j}$ ,  $j = 1, \dots, p-1$ , are  $(k \times k)$  matrices.  $\Phi_{2,j}$  and  $\Phi_{4,j}$ ,  $j = 0, \dots, p-1$ , are  $(k \times m)$  matrices and  $\mu_i$ ,  $i = 1, 2$ , are  $(k \times 1)$  vectors.  $G(s_t)$  is the transition function, assumed to be continuous and bounded between zero and one. The STVECM framework can be considered as a regime-switching model which allows for two regimes,  $G(s_t) = 0$  and  $G(s_t) = 1$ , respectively, where the transition from one to the other regime occurs in a smooth way. There are three popular choices for  $G(s_t)$ . The first one is the ‘logistic’ function:

$$G(s_t; \gamma, c) = \{1 + \exp[-\gamma(s_t - c)/\sigma(s_t)]\}^{-1}, \gamma > 0, \quad (2a)$$

where  $\sigma(s_t)$  is the sample standard deviation of  $s_t$ . This model assumes asymmetric adjustment to positive and negative deviations of  $s_t$  relative to a parameter  $c$ . The latter is the threshold between the two regimes;  $G(s_t)$  changes monotonically from 0 to 1 as  $s_t$  increases, and takes the value of

$G(s_t) = 0.5$  at  $s_t = c$ . The parameter  $\gamma$  determines the speed of the transition from one regime to the other. When  $\gamma \rightarrow 0$ , the ‘logistic’ function equals a constant (i.e. 0.5), and when  $\gamma \rightarrow +\infty$ , the transition from  $G(s_t) = 0$  to  $G(s_t) = 1$  is almost instantaneous at  $s_t = c$ .

The second choice is the ‘exponential’ function:

$$G(s_t; \gamma, c) = 1 - \exp\{-\gamma(s_t - c)^2 / \sigma^2(s_t)\}, \gamma > 0. \quad (2b)$$

This model assumes asymmetric adjustment to small and large absolute values of  $s_t$ . The third choice derives from the fact that the ‘exponential’ model above becomes linear if either  $\gamma \rightarrow 0$  or  $\gamma \rightarrow +\infty$ . This can be avoided by setting  $G(s_t)$  equal to the ‘quadratic logistic’ function:

$$G(s_t; \gamma, c_1, c_2) = \{1 + \exp[-\gamma(s_t - c_1)(s_t - c_2) / \sigma^2(s_t)]\}^{-1}, \gamma > 0, \quad (2c)$$

as proposed by Jansen and Teräsvirta (1996). In this case, if  $\gamma \rightarrow 0$ , the model becomes linear, whereas if  $\gamma \rightarrow +\infty$ ,  $G(s_t)$  is equal to 1 for  $s_t < c_1$  and  $s_t > c_2$ , and equal to 0 when  $c_2 < s_t < c_1$ .

In this paper we assume that the possible candidates for the transition variable  $s_t$  are the  $r$  cointegrating relationships in  $z_{t-1} = \beta'[y_{t-1}', x_{t-1}']'$ . More specifically, the next section estimates one cointegrating vector which is identified as a long-run real exchange rate equation. In order to model non-linear deviations of the UK real exchange rate from its long-run level we confine our attention to the transition function (2c) rather than models (2a) and (2b) for two reasons. First, model (2a) allows for asymmetric adjustment of positive and negative deviations of the disequilibrium error relative to a single threshold  $c$ , which does not look plausible on economic grounds (see also the discussion in Michael *et al.*, 1997). Second, the ‘exponential’ model (2b) has the drawback that it becomes linear for very small or very large values of the  $\gamma$  parameter as

discussed above. For these reasons, we proceed by adopting model (2c) as the most plausible transition function. From an economic point of view, model (2c) is particularly attractive as it implies the existence of an interval band  $(c_1, c_2)$  outside which there is a strong tendency for the real exchange rate to revert to its equilibrium value. Dumas (1992) justifies the existence of this band in terms of the costs of trading goods; to the extent that deviations of the real exchange rate from its long-run level are small relative to the costs of trading, these deviations are left uncorrected (see also the discussion in Michael *et al.*, 1997, and the references therein).

A test of linearity in model (1) using the transition function (2c), is a test of the null hypothesis  $H_0: \gamma = 0$  against the alternative  $H_1: \gamma > 0$ . By taking a first-order Taylor approximation of  $G(s_t)$  around  $\gamma = 0$ , the test can be done within the reparameterised model (see e.g. the discussion in Saikkonen and Luukkonen, 1988):

$$\begin{aligned} \Delta y_t = & M_0 + A_0 z_{t-1} + \sum_{j=1}^{p-1} B_{0,j} \Delta y_{t-j} + \sum_{j=0}^{p-1} B_{1,j} \Delta x_{t-j} + A_1 z_{t-1} s_t + \sum_{j=1}^{p-1} B_{2,j} \Delta y_{t-j} s_t + \\ & \sum_{j=0}^{p-1} B_{3,j} \Delta x_{t-j} s_t + A_2 z_{t-1} s_t^2 + \sum_{j=1}^{p-1} B_{4,j} \Delta y_{t-j} s_t^2 + \sum_{j=0}^{p-1} B_{5,j} \Delta x_{t-j} s_t^2 + e_t, \end{aligned} \quad (3)$$

where  $e_t$  are the original errors  $\varepsilon_t$  plus the error arising from the Taylor approximation. Model (3) is a linear VECM augmented by additional cross-product regressors due to the Taylor expansion. Here, the null hypothesis of linearity is  $H_0' : A_1 = A_2 = B_{2,j} = B_{3,j} = B_{4,j} = B_{5,j} = 0$ , where  $j = 1, \dots, p-1$  for the  $B_{2,j}$  and  $B_{4,j}$  matrices and  $j = 0, \dots, p-1$  for the  $B_{3,j}$  and  $B_{5,j}$  matrices. For each equation in the VECM, this is a standard variable addition Lagrange Multiplier (LM) which follows asymptotically the  $\chi^2$  distribution with  $2r + 2k(p-1) + 2mp$  degrees of freedom. In small samples, the  $\chi^2$  test may be heavily oversized. Therefore, it may be preferable to use an  $F$  version of the test. Both the  $\chi^2$  and  $F$  versions of the LM statistic are equation specific tests for linearity

which are computed from an auxiliary regression of the residuals from each equation in the linear VECM on all variables entering model (3). To test the null hypothesis of linearity in all equations simultaneously, we need a system-wide test. Following Weise (1999), define  $\Omega_0$  and  $\Omega_1$  as the estimated variance-covariance residual matrices from the linear VECM and the augmented model (3), respectively. The appropriate log-likelihood system-wide test statistic is given by  $LR = T\{\log|\Omega_0| - \log|\Omega_1|\}$ , where  $T$  is the size of the sample. Under the null hypothesis of linearity, the test follows asymptotically the  $\chi^2$  distribution with  $2rk + 2k^2(p - 1) + 2kmp$  degrees of freedom. The equation specific  $LM$  tests and the system wide  $LR$  test are run for all possible  $s$ , candidates, that is, all cointegrating relationships. The decision rule is to select as the appropriate transition variable the cointegrating relationship for which the  $p$ -value of the test statistic is the smallest one.<sup>1</sup>

### 3. Econometric specification of a real exchange rate model

The theoretical framework discussed above will now be tested on a small model of the UK real exchange rate. In an earlier paper, Alogoskoufis (1990) introduces a model with traded ( $T$ ) and non-traded ( $NT$ ) goods.<sup>2</sup> The model assumes perfect competition in the  $T$  sector with firms producing according to a two-level CES production function which is separable into capital, labour and imported oil. For the  $NT$  sector, the model assumes profit maximising monopolistic competitive firms. The relative price of tradables to the price of domestic output,  $p_T - p$  is derived as:

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<sup>1</sup> An extension of the Saikkonen and Luukkonen (1988) linearity tests involves a second-order Taylor approximation of the transition function as suggested by Escibano and Jordá (1999). This involves adding cubic and fourth power terms in model (3), which is hardly practical to implement since we are faced with a small sample size. Further, as van Dijk *et al.* (2001) point out, neither one of the tests in Saikkonen and Luukkonen (1988) or Escibano and Jordá (1999) dominates in terms of power.

<sup>2</sup> For other versions of price models with traded and non-traded goods see e.g. Martin (1997) and the references therein.



$$p_T - p = -(1 - \tau)(w - p_T) + \tau \left( \frac{1 - \pi_1}{\pi_1} \right) (p_O - p_T), \quad (4)$$

where  $\pi_1$  ( $0 < \pi_1 < 1$ ) is the share of value added in gross output,  $\tau$  is the share of tradables in total output,  $(w - p_T)$  refers to real product wages in the tradables sector and  $(p_O - p_T)$  is the relative price of imported oil. All variables are in logs. Assuming that the UK is a small open economy, the price of domestic tradables  $p_T$  can be proxied by  $p_T = p_T^* + e$ , where  $p_T^*$  is the price index of UK imports in \$, and  $e$  is the average £/\$ exchange rate. In this case, equation (4) is a measure of the real exchange rate as a negative function of  $w - p_T$  and a positive function of  $p_O - p_T$ . An increase in  $p_T - p$  is equivalent to a real depreciation or an improvement in the real competitiveness of the domestic economy.

Following the notation in Section 2 of the paper, our model uses a set of  $k = 3$  endogenous variables:

$$y = [p_T - p, w - p_T, u]', \quad (5)$$

conditioning on  $x = p_O - p_T$ , that is,  $m = 1$  exogenous variable. We use quarterly seasonally adjusted UK data over the period 1973(1)-2000(1). The endogenous variables refer to the real exchange rate,  $p_T - p$ , real product wages,  $w - p_T$ , in the manufacturing sector (as a proxy for tradables), and the unemployment rate,  $u$  (all variables are in logs; for more details see the Data appendix).<sup>3</sup> Firms are assumed to take the real price of oil,  $p_O - p_T$ , as given and therefore we impose exogeneity of this variable, which may improve the statistical properties of the system (see the discussion in Hansen and Juselius, 1994).

What we are interested in, is to see whether deviations of the real exchange rate equation from

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<sup>3</sup> As discussed in the Appendix, we use wage costs rather than wage costs per unit of output in our empirical results. That said, use of unit wage costs does not make any qualitative difference to the results reported below.

its long-run equilibrium level given by (4) affect the short-run dynamics of system (5) in a non-linear way. The unemployment rate,  $u$ , is included in the system for two reasons. First, we would like to examine its relationship with wages. Second, and more important, we would like to test whether unemployment is affected by the discrepancy between the real exchange rate and its implied long-run equilibrium relationship, that is, the magnitude of the real exchange rate misalignment. A negative effect would suggest that when the real exchange rate is above its long-run equilibrium level (i.e. the real exchange rate is undervalued), firms respond to an improvement in domestic competitiveness by increasing their demand for labour and as a result, unemployment falls.<sup>4</sup>

## 4. Empirical results

### 4.1 Long-run behaviour

Figure 1 plots the logs of the levels and the first differences of the  $p_T - p$ ,  $w - p_T$ ,  $p_O - p_T$  and  $u$  series. Preliminary analysis using the ADF unit root tests suggested that all series are  $I(1)$  in levels and  $I(0)$  in first differences. Before testing for cointegration, we estimate the linear VECM in levels using a lag length of  $p = 4$  and allowing for a drift parameter to enter the VECM unrestrictedly.<sup>5</sup> Table 1 reports the eigenvalues,  $\lambda_i$ , and the  $\lambda$ -max and trace statistic tests for cointegration (see Johansen, 1988). The 95 percent critical values are derived from Microfit 4.0

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<sup>4</sup> Obviously, some other variables like productivity or tax rates can affect wages or unemployment. Extending the information set to this direction is not pursued here, as we are primarily interested in discussing the non-linear behaviour of the real exchange rate equation. Furthermore, use of a larger information set is impractical because the number of estimated coefficients in the linearity tests and the STVECM rises considerably relative to the number of estimated coefficients in the linear VECM. For these reasons we settle for a relatively small baseline system.

<sup>5</sup> The lag length is selected by starting with 5 lags on each variable, and sequentially testing down using an F-test. The three endogenous equations  $y = [p_T - p, w - p_T, u]'$  in the linear VECM pass the Autocorrelation test (of up to order 5) and the 4th order Autoregressive Conditional Heteroscedasticity test and Heteroscedasticity. They fail Normality and Heteroscedasticity. Detailed diagnostic tests for the linear and non-linear versions of the short-run equations are reported below.

(Pesaran and Pesaran, 1997) which uses the correct critical values in the case of a linear system with both endogenous and exogenous  $I(1)$  variables (for more details see Pesaran *et al.*, 2000). Both the  $\lambda$ -max and *trace* statistics indicate the existence of  $r = 1$  cointegrating vector. For exact identification, we normalise the estimated vector on the real exchange rate,  $p_T - p$ . Then we test one over-identifying restriction, that is, long-run exclusion of the unemployment rate,  $u$ . The restriction is accepted as it calculates  $\chi^2(1) = 2.73$  ( $p$ -value = 0.10) and the resulting cointegrating vector is:

$$p_T - p = \begin{matrix} -0.457 & +0.182 \\ (0.030) & (0.111) \end{matrix} \begin{matrix} (w - p_T) \\ (p_o - p_T) \end{matrix}$$

where standard errors are given in parentheses below the estimated coefficients. The estimated cointegrating relationship looks like the theoretical real exchange rate equation (4) with the share of traded goods in total output (i.e.  $\tau$ ) estimated at 54.3 percent and the share of value added in gross output (i.e.  $\pi_1$ ) estimated at 74.8 percent (the latter is derived from  $\tau(1 - \pi_1) / \pi_1 = 0.182$ ).<sup>6 7</sup> Figure 2 plots the mean-corrected deviations from the estimated relationship. Movements of the disequilibrium error above (below) the zero line are associated with an undervalued (overvalued) real exchange rate. We discuss this issue further in section 5 of the paper.

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<sup>6</sup> Using annual data over the 1952-1985 period, Alogoskoufis (1990) estimates  $\tau$  between 31 percent and 39 percent, and  $\pi_1$  at 92 percent. However, he points out that his estimates for  $\tau$  are implausibly low. Our estimate for  $\pi_1$  at 74.8 percent is much closer to  $\pi_1$  at 71.7 percent in Bruno and Sachs (1982). Their estimate is based on a system of factor-price frontier, output supply, and labour demand equations using annual data over the 1956-1978 period.

<sup>7</sup> Chaudhuri and Daniel (1998) adopt the Engle-Granger two-step procedure to test for cointegration in a bivariate model involving real exchange rates and real oil prices for sixteen OECD countries. Using monthly data over the 1973(1)-1996(2) period, they find that the UK real exchange rate cointegrates with the real price of oil. The coefficient on the real price of oil is estimated at 0.389.

#### 4.2 Linearity testing and short-run estimates

Having estimated the long-run real exchange rate equation, we test for linearity in model (3) using the estimated cointegrating vector  $CV_{t-1}$  as the possible transition variable  $s_t$ . Linearity tests are run for a different number of lags of the transition variable  $s_{t-d} = CV_{t-d}$  (namely  $d = 1, 2, 3,$  and  $4$  lags). Then the appropriate lag is selected as the one for which the linearity test is most strongly rejected. We report bootstrapped  $p$ -values instead of asymptotic  $p$ -values although our results are not particularly sensitive to the above choice. To compute the bootstrapped  $p$ -values of the equation specific  $F$  tests and the system wide  $LR$  test reported in Table 2, we followed closely Weise (1999). First, we estimated the linear VECM equations. To control for the presence of heteroscedasticity in the linear models, the VECM residuals were regressed on all RHS variables entering the linear VECM as well as their squares, and the original residuals were transformed using the estimated coefficients from this auxiliary regression. Draws were taken from the transformed residuals and one thousand artificial data series were constructed. For each of these artificial series,  $F$  and  $LR$  statistics were constructed and then compared to the corresponding statistics from the actual data. The bootstrapped  $p$ -values were derived as the number of times the  $F$  and  $LR$  statistics from the artificial data exceeded the corresponding statistics from the actual data, divided by one thousand.

According to the results in Table 2, linearity is mostly rejected for  $CV_{t-1}$ . Using the disequilibrium error  $CV_{t-1}$  in the ‘quadratic logistic’ function (2c), we therefore proceed by estimating the non-linear short-run  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  equations.<sup>8</sup> Before estimating

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<sup>8</sup> In that respect, our paper is different from the studies of Michael *et al.* (1997) and Sarantis (1999) who estimate STAR real exchange rate models for a number of countries. Both studies look at the possibility of non-linear adjustment within a univariate rather than a multivariate framework. Using the transition function (2b), Michael *et al.* (1997) apply a STAR model to the residuals of a Purchasing Power Parity (PPP) relationship. They select lagged values of the PPP residuals as possible transition variable candidates. Using the transition functions (2a) and (2b), Sarantis (1999) applies a STAR model to the first differences of the real exchange rates. He selects lagged real exchange rate values as possible transition variable candidates.

the non-linear models, it is worth mentioning that Granger and Teräsvirta (1993) and Teräsvirta (1994) stress particular problems like slow convergence or overestimation associated with estimates of the  $\gamma$  parameter. For this reason, we follow their suggestion in scaling the ‘quadratic logistic’ function (2c) by dividing it by the variance of the transition variable  $\sigma^2(CV_{t-1})$  (which equals 0.003), so that  $\gamma$  becomes a scale-free parameter. Based on this scaling, we use  $\gamma = 1$  as a starting value and values of  $CV_{t-1}$  close to its minimum (which equals  $-0.151$ ) and maximum (which equals  $0.106$ ) as starting values for the parameters  $c_1$  and  $c_2$ , respectively. The estimates of the linear equations for  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  are used as starting values for the remaining parameters in the STVECM equations (1).

For comparison reasons, we report both the linear and the non-linear versions of the estimated equations. Tables 3A to 3C report the OLS estimates of the parsimonious linear models, whereas Tables 4 to 6 report the non-linear least squares (NLS) estimates for the parsimonious STVECM equations (1) adopting the quadratic logistic transition function (2c). The estimations have been done in PcGive 9.0 (Hendry and Doornik, 1997).<sup>9</sup>

The main parameters of interest in the non-linear models are the estimated values of the threshold parameters  $c_1$  and  $c_2$ , and the speed of adjustment,  $\gamma$ . The  $c_1$  and  $c_2$  estimates reported in Tables 4 to 6 are statistically significant in all models. The  $c_1$  and  $c_2$  estimates indicate the existence of two regimes for the  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  equations; one characterised by large deviations of the real exchange rate from its long-run equilibrium and an alternative one which is characterised by small real exchange rate deviations from its equilibrium level. The economic implications of these results will be discussed in the following section. The estimates

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<sup>9</sup> One could also argue in favour of a structural rather than a reduced form model by testing the significance of current  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$ , and  $\Delta u_t$  effects in the estimated equations. However, these effects were insignificant.

of the  $\gamma$  parameter are rather high for all models indicating that the speed of the transition from  $G(s_t; \gamma, c_1, c_2) = 0$  to  $G(s_t; \gamma, c_1, c_2) = 1$  is rapid at the estimated thresholds  $c_1$  and  $c_2$ . Notice, however, the rather high standard error of the  $\gamma$  estimates. Teräsvirta (1994) and van Dijk *et al.* (2001) point out that this should not be interpreted as evidence of weak non-linearity. Accurate estimation of  $\gamma$  is not always feasible, as it requires many observations in the immediate neighborhood of the threshold parameters  $c_1$  and  $c_2$ . Further, large changes in  $\gamma$  have only a small effect on the shape of the transition function implying that high accuracy in estimating  $\gamma$  is not necessary (see the discussion in van Dijk *et al.*, 2001).

From Tables 3 to 6 one can notice a large improvement in the diagnostic tests of the non-linear relative to the linear models. The error variance ratio of the non-linear relative to the linear models (i.e.  $s^2_{NL}/s^2_L$ ) is less than one, indicating that the non-linear models have a better fit. In particular, the  $s^2_{NL}/s^2_L$  ratio shows a reduction in the residual variances of the non-linear compared to the linear models which ranges from around 16 percent for the  $\Delta(p_T - p)_t$  and  $\Delta u_t$  equations in Table 4 and Table 6, respectively, to around 30 percent for the  $\Delta(w - p_T)_t$  model in Table 5. In addition, the non-linear specification of all three models captures the heteroscedastic and most of the normality failures that are present in the corresponding linear models.

## 5. Discussion of the results

Looking at the linear short-run equations first, one can notice that the cointegrating vector ( $CV_{t-1}$ ) enters with the correct sign in the  $\Delta(p_T - p)_t$  equation (see Table 3A). The  $CV_{t-1}$  effect (i.e.  $-0.107$ ) suggests a slow adjustment to disequilibrium deviations of the real exchange rate from its optimal level determined by the long-run relationship. Bearing in mind that the real exchange rate equation captures aspects of the real competitiveness of the domestic economy that depend on

conditions prevailing in the production sector, this sluggishness could reflect rigidities in the functioning of the product market due to product heterogeneity, government imposed barriers to trade, or labour market inflexibility distorting the adjustment of wages.<sup>10</sup> The disequilibrium error also affects negatively the  $\Delta u_t$  equation (i.e. coefficient on  $CV_{t-1}$  equals  $-0.236$  in Table 3C). This result implies that when the real exchange rate is above its equilibrium level, that is, undervalued, unemployment falls as firms respond to an improvement in domestic competitiveness by increasing their demand for labour. We also report a short-run negative effect of past real wage growth (i.e.  $\Delta(w - p_T)_{t-1}$ ) in the  $\Delta u_t$  equation. The cointegrating vector has a weak positive effect in the  $\Delta(w - p_T)_t$  equation (i.e. coefficient on  $CV_{t-1}$  equals  $0.094$  in Table 3B). Hence, there seems to be some weak evidence that when the real exchange rate is undervalued, workers respond to an improvement in domestic competitiveness by demanding and getting higher wages. The estimates in Table 3B also suggest a significant effect from past changes in unemployment (i.e.  $\Delta u_{t-1}$ ,  $\Delta u_{t-2}$ , and  $\Delta u_{t-3}$ ) in the  $\Delta(w - p_T)_t$  equation but these effects seem to cancel each other out.<sup>11</sup>

The NLS estimates suggest that all three  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  equations exhibit a regime-switching behaviour according to the variation of the disequilibrium error. Consider first the  $\Delta(p_T - p)_t$  equation in Table 4. The estimate of the disequilibrium error in the second regime (i.e. coefficient  $\alpha_2$  is equal to  $-0.143$  when  $G(s_t; \gamma, c_1, c_2) = 1$ ) is higher than that of the disequilibrium error in the first regime (i.e. coefficient  $\alpha_1$  is equal to  $-0.094$  when  $G(s_t; \gamma, c_1, c_2) = 0$ ). This implies that when the real exchange rate exceeds an estimated interval band of  $(c_1, c_2) = (-0.096, 0.078)$ , the short-run real exchange rate adjusts faster.

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<sup>10</sup> See also the discussion in e.g. Johansen and Juselius (1992) and Pesaran and Shin (1996) in the context of a system involving the UK effective exchange rate, UK interest rate, UK prices, foreign prices and foreign interest rate.

<sup>11</sup> Among other studies, Manning (1993) using annual data over the 1956-1987 period, reports a negative effect from real wages on unemployment as well as a negative effect from unemployment on real wages. However, his model uses the level of  $u$  rather than  $\Delta u$  which is used here.

From Table 6 one can see that short-run unemployment  $\Delta u_t$  reacts fast to the disequilibrium error (i.e. coefficient  $\alpha_2$  is equal to  $-0.291$ ) only in the second regime (i.e. when  $G(s_t; \gamma, c_1, c_2) = 1$ ), that is, when the disequilibrium error exceeds an estimated interval band of  $(c_1, c_2) = (-0.040, 0.052)$ . In addition, the estimated interval band  $(c_1, c_2)$  for the short-run unemployment rate is much narrower compared to that for the short-run real exchange rate. Taking into account that our sample covers a period of floating exchange rates (with the UK joining the Exchange Rate Mechanism only between 1990 and 1992), it is reasonable to expect that the short-run real exchange rate  $\Delta(p_T - p)_t$  will adjust faster when the cointegrating vector  $CV_{t-1}$  is outside a rather wide interval band of thresholds. To the extent that the real exchange rate equation reflects monetary and more generally economic policy considerations, the significant effect of the cointegrating vector in the short-run unemployment equation implies that unemployment can be targeted by economic policy. Further, the lower estimates of  $c_1$  and  $c_2$  for the  $\Delta u_t$  equation suggest that if economic authorities want to avoid large swings in unemployment then they should be prepared to keep real exchange rate movements within a narrow interval band of thresholds.<sup>12</sup>

Our results in Table 5 suggest that short-run wages  $\Delta(w - p_T)_t$  are affected by real exchange rate fluctuations within an estimated interval band of  $(c_1, c_2) = (-0.070, 0.103)$  (i.e. coefficient  $\alpha_1$  is equal to  $0.102$  in the first regime  $G(s_t; \gamma, c_1, c_2) = 0$ ). Contrary to our results for the  $\Delta(p_T - p)_t$  and  $\Delta u_t$  equations, we could not find any significant effect from the disequilibrium error on short-run wages outside the estimated band of thresholds. This is rather surprising, as we would expect the impact of the disequilibrium error on the short-run dynamics to be more evident when

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<sup>12</sup> That said, it is worth pointing out that policy makers in the UK were never able to control the exchange rate. Describing the main characteristics of macroeconomic policies in the UK, Andrew Britton, the former director of the National Institute of Economic and Social Research, comments: “Attempts to use the exchange rate as a policy instrument misfired; attempts to control it failed; attempts to ignore it were no more successful. The authorities never really got on top of the situation at all.” (Britton, 1991, p. 298).



deviations of the real exchange rate from its long-run level exceed the estimated interval band. This finding probably has to do with the rather wide interval band estimates and in particular the  $c_2$  estimate, which is practically equal to the maximum value of  $CV_{t-1}$  (i.e. 0.106).

The relationship between the occurrence of a regime and the disequilibrium error is depicted in Figure 3, which plots the values of the transition function against  $CV_{t-1}$  for the  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  equations. As discussed above,  $G(s_t; \gamma, c_1, c_2) = 0$  and  $G(s_t; \gamma, c_1, c_2) = 1$  are related to small and large deviations respectively of the real exchange rate relative to a band of thresholds. In addition, this Figure helps clarify the discussion about the speed of transition between the two regimes. One can see that the transition from one regime to the other is rapid, as the estimates of  $\gamma$  are rather high for all models.

Figure 4 plots the estimated transition function for each model against time in order to illustrate the succession of regimes over the sample period. From Figure 4A, the 1980-1982 and 1995-1997 periods are classified into the second regime of real exchange rate deviations outside the estimated interval bands. One can notice from Figure 2 that during the 1980-1982 period, the estimated transition functions pick a highly overvalued real exchange rate, whereas during the 1995-1997 period, they pick a highly undervalued real exchange rate which then began reverting to its long-run level. Figure 4C shows that switches from one regime to the other are particularly active for the unemployment equation where a much narrower interval band was estimated. It is notable that the first period, which captures the 1980-1981 economic recession, follows the second OPEC oil price hike (an increase in oil prices of around 15 percent in June 1979) and coincides with important changes in economic policies following the election of the Thatcher government in May 1979. In particular, 1979 saw the abolition of exchange rate controls, which was not aimed at any particular effect on the exchange rate, as well as public spending cuts and an increase in indirect taxation. The new government encouraged the use of cheaper labour,

especially female labour, which led to more part-time employment. At the same time, a very tight monetary policy aiming at a rapid decrease in the rate of inflation, led to a more overvalued real exchange rate (see Figure 2), a rapid increase in unemployment (see Figure 1) and a severe recession (see e.g. Britton, 1991; Mizon, 1995). Taking into account the slow adjustment of the real exchange rate reported in the previous section of the paper, it is not surprising that after the UK's exit from the ERM in September 1992, the real exchange rate experienced a path of persistent depreciation, which peaked between 1995 and 1997 (see Figure 2). Indeed, this is the non-linearity in the short-run real exchange rate captured by Figure 4A.

## **6. Conclusions**

In this paper we discussed linearity testing within a STVECM framework. The theoretical framework was illustrated by looking at non-linearities in the case of the UK real exchange rate. After estimating a long-run real exchange rate equation as part of a small system involving real product wages, the unemployment rate and the real price of oil, we found evidence that deviations of the real exchange rate equation from its long-run equilibrium level affect in a non-linear way not only the short-run real exchange rate equation, but also the short-run unemployment and wage equations. According to our estimates, the short-run real exchange rate adjusts faster when the cointegrating relationship is outside a wide interval band of thresholds. This is not surprising as our sample covers a period of floating exchange rates. On the other hand, the short-run unemployment rate adjusts fast when the cointegrating relationship is outside a narrower interval band. To the extent that the real exchange rate equation reflects economic policy considerations, our results suggest that policy makers should aim at a narrow band for the real exchange rate if they want to avoid large swings in unemployment.

Bearing in mind that the development of multivariate non-linear models is still at an early

stage, our prime intention in this paper was to examine non-linearities within such a framework using the UK real exchange rate as an example rather than give definite answers about its non-linear adjustment. These answers are going to be the subject of our future research.

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### **Data appendix**

$p$  : GDP value added price deflator (1995 = 100). *Source: Economic Trends Annual Supplement (ETAS).*

$p_T = p_T^* + e$ : the price of domestic tradables, where  $p_T^*$  is the \$ price index of UK imports (1990 = 100) and  $e$  is the average £/\$ exchange rate. *Source: IMF, International Financial Statistics.*

$w = w_{NET} + t_1$ : average product wages in the manufacturing sector (1990=100), where  $w_{NET}$  refers to average weekly wages in manufacturing (net of employers' taxes) and  $t_1$  is the tax rate on labour paid by employers, constructed as: (employers' contributions) / (total wage bill). *Source: ETAS.*

$p_o$  : price index of materials and fuels purchased by manufacturers (1995 = 100). *Source: ETAS.*

$u$  : UK unemployment rate. *Source: ETAS.* All variables are in logs.

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Table 1  
Eigenvalues, test statistics and critical values

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$\lambda_i$	$\lambda$ -max				trace			
	H <sub>0</sub>	H <sub>1</sub>	Stat.	95%	H <sub>0</sub>	H <sub>1</sub>	Stat.	95%
0.23	r = 0	r = 1	26.82	24.59	r = 0	r ≥ 1	44.65	38.93
0.14	r ≤ 1	r = 2	15.67	18.06	r ≤ 1	r ≥ 2	17.83	23.32
0.02	r ≤ 2	r = 3	2.16	11.47	r ≤ 2	r = 3	2.16	11.47

Notes:  $r$  denotes the number of cointegration vectors. Critical values are from Microfit 4.0 (Pesaran and Pesaran, 1997).

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Table 2  
Linearity tests

Transition variable	Lagrange Multiplier $F$ statistics for:			
	$\Delta(p_T - p)$ model	$\Delta(w - p_T)$ model	$\Delta u$ model	System wide test $LR$
$CV_{t-1}$	1.111 (0.375)	1.888 (0.036)	2.082 (0.008)	103.352 (0.009)
$CV_{t-2}$	0.855 (0.684)	1.361 (0.180)	1.833 (0.020)	96.414 (0.029)
$CV_{t-3}$	0.973 (0.511)	1.914 (0.027)	1.173 (0.311)	89.446 (0.067)
$CV_{t-4}$	0.933 (0.555)	2.103 (0.016)	1.044 (0.437)	85.701 (0.094)

Notes: Bootstrapped  $p$ -values in parentheses. The  $p$ -values for the equation specific Lagrange Multiplier  $F$  statistics and the system wide  $LR$  test statistic are derived from bootstrapping with one thousand replications.  $CV$  is the transition variable:  $CV = p_T - p + 0.457 (w - p_T) - 0.182 (p_O - p_T)$ , in mean-corrected form. The null hypothesis is linearity. The alternative hypothesis is the STVECM representation.



Table 3  
Estimated linear models

Panel A: Linear  $\Delta(p_T - p)_t$  model

$$\begin{aligned} \Delta(p_T - p)_t = & -0.001 & -0.107 & +0.277 & -0.210 \\ & (0.002) & (0.044) & (0.171) & (0.158) \\ & -0.178 & +0.401 & -0.127 \\ & (0.103) & (0.104) & (0.106) \end{aligned} \begin{array}{l} CV_{t-1} \\ \Delta(p_T - p)_{t-1} \\ \Delta(w - p_T)_{t-1} \\ \Delta(p_O - p_T)_t \\ \Delta(p_O - p_T)_{t-1} \\ \Delta(p_O - p_T)_{t-2} \end{array}$$

$s_L = 0.023$ , AR(5) = 0.39[0.856], ARCH(4) = 0.11[0.980], HET = 1.99[0.035], NORM(2) = 21.7[0.000]

Panel B: Linear  $\Delta(w - p_T)_t$  model

$$\begin{aligned} \Delta(w - p_T)_t = & 0.008 & +0.094 & -0.415 & +0.188 \\ & (0.002) & (0.060) & (0.097) & (0.092) \\ & -0.263 & +0.154 & +0.155 & -0.396 \\ & (0.120) & (0.083) & (0.113) & (0.116) \end{aligned} \begin{array}{l} CV_{t-1} \\ \Delta(p_T - p)_{t-1} \\ \Delta u_{t-1} \\ \Delta u_{t-2} \\ \Delta u_{t-3} \\ \Delta(p_O - p_T)_t \\ \Delta(p_O - p_T)_{t-1} \end{array}$$

$s_L = 0.025$ , AR(5) = 0.28[0.923], ARCH(4) = 0.60[0.662], HET = 4.60[0.000], NORM(2) = 34.1[0.000]

Panel C: Linear  $\Delta u_t$  model

$$\begin{aligned} \Delta u_t = & 0.004 & -0.236 & -0.350 & -0.347 \\ & (0.002) & (0.055) & (0.187) & (0.170) \\ & +0.897 & -0.203 & -0.256 \\ & (0.063) & (0.058) & (0.107) \end{aligned} \begin{array}{l} CV_{t-1} \\ \Delta(p_T - p)_{t-1} \\ \Delta(w - p_T)_{t-1} \\ \Delta u_{t-1} \\ \Delta u_{t-3} \\ \Delta(p_O - p_T)_t \end{array}$$

$s_L = 0.024$ , AR(5) = 1.10[0.364], ARCH(4) = 1.62[0.177], HET = 2.29[0.014], NORM(2) = 7.99[0.018]

Notes: Standard errors are given in parentheses below the estimates.  $s_L$ : standard error of the linear regression. AR(5): F-test for up to 5th order serial correlation. ARCH(4): 4th order Autoregressive Conditional Heteroscedasticity F-test. HET: F-test for Heteroscedasticity. NORM(2): Chi-square test for normality. Numbers in square brackets are the  $p$ -values of the test statistics.

Table 4

Estimated non-linear  $\Delta(p_T - p)_t$  model

The Table reports the NLS estimates of the following STVECM equation:

$$\begin{aligned} \Delta(p_T - p)_t = & (\mu_1 + \alpha_1 CV_{t-1} + \phi_{1,1} \Delta(p_T - p)_{t-1} + \phi_{2,1} \Delta(p_O - p_T)_{t-1}) (1 - G(CV_{t-1}; \gamma, c_1, c_2)) \\ & + (\mu_2 + \alpha_2 CV_{t-1} + \phi_{3,1} \Delta(w - p_T)_{t-1} + \phi_{4,1} \Delta(p_O - p_T)_t + \phi_{4,2} \Delta(p_O - p_T)_{t-1} \\ & + \phi_{4,3} \Delta(p_O - p_T)_{t-2}) G(CV_{t-1}; \gamma, c_1, c_2) \end{aligned}$$

$$\text{where } G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-\gamma (CV_{t-1} - c_1) (CV_{t-1} - c_2) / \sigma^2(CV_{t-1})]\}^{-1},$$

is the ‘quadratic logistic’ transition function, with  $CV_{t-1}$  as the transition variable. Values of 0 and 1 of the transition function are associated with the two alternative regimes. The  $\Delta(p_T - p)_t$  dynamics in the first regime, when  $G(CV_{t-1}; \gamma, c_1, c_2) = 0$ , are:

$$\Delta(p_T - p)_t = (\mu_1 + \alpha_1 CV_{t-1} + \phi_{1,1} \Delta(p_T - p)_{t-1} + \phi_{2,1} \Delta(p_O - p_T)_{t-1}) (1 - G(CV_{t-1}; \gamma, c_1, c_2)).$$

In the second regime, when  $G(CV_{t-1}; \gamma, c_1, c_2) = 1$ , its dynamics are:

$$\begin{aligned} \Delta(p_T - p)_t = & (\mu_2 + \alpha_2 CV_{t-1} + \phi_{3,1} \Delta(w - p_T)_{t-1} + \phi_{4,1} \Delta(p_O - p_T)_t + \phi_{4,2} \Delta(p_O - p_T)_{t-1} \\ & + \phi_{4,3} \Delta(p_O - p_T)_{t-2}) G(CV_{t-1}; \gamma, c_1, c_2) \end{aligned}$$

For intermediate values of  $G(CV_{t-1}; \gamma, c_1, c_2)$ , i.e.  $0 < G(CV_{t-1}; \gamma, c_1, c_2) < 1$ ,  $\Delta(p_T - p)_t$  dynamics are a weighted average of the two equations. The speed of transition between the two regimes is determined by the parameter  $\gamma$ .

$$\begin{aligned} \Delta(p_T - p)_t = & \begin{pmatrix} -0.003 & -0.094 CV_{t-1} & +0.387 \Delta(p_T - p)_{t-1} \\ (0.002) & (0.059) & (0.082) \end{pmatrix} \\ & + 0.247 \Delta(p_O - p_T)_{t-1} (1 - G(CV_{t-1}; \gamma, c_1, c_2)) \\ & (0.096) \\ & \begin{pmatrix} +0.010 & -0.143 CV_{t-1} & -1.364 \Delta(w - p_T)_{t-1} & -0.422 \Delta(p_O - p_T)_t \\ (0.007) & (0.092) & (0.362) & (0.372) \end{pmatrix} \\ & + 0.694 \Delta(p_O - p_T)_{t-1} - 0.434 \Delta(p_O - p_T)_{t-2} G(CV_{t-1}; \gamma, c_1, c_2) \\ & (0.419) \quad (0.432) \end{aligned}$$

where

$$G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-7.016(CV_{t-1} + 0.096) (CV_{t-1} - 0.078) / \sigma^2(CV_{t-1})]\}^{-1}$$

$$(3.923) \quad (0.003) \quad (0.006)$$

$$s_{NL} = 0.021, \quad s_{NL}^2 / s_L^2 = 0.834, \quad \text{AR}(5) = 2.85[0.020], \quad \text{ARCH}(4) = 0.23[0.917], \quad \text{HET} = 0.29[0.999],$$

$$\text{NORM}(2) = 8.24[0.016]$$

Notes: Standard errors are given in parentheses below the estimates.  $s_{NL}$ : standard error of the non-linear regression. The diagnostic tests are discussed in the notes of Table 3.

Table 5

Estimated non-linear  $\Delta(w - p_T)_t$  model

The Table reports the NLS estimates of the following STVECM equation:

$$\begin{aligned} \Delta(w - p_T)_t = & (\mu_1 + \alpha_1 CV_{t-1} + \phi_{1,1} \Delta(p_T - p)_{t-1} + \phi_{2,1} \Delta(p_O - p_T)_{t-1}) (1 - G(CV_{t-1}; \gamma, c_1, c_2)) \\ & + (\mu_2 + \phi_{3,1} \Delta(p_T - p)_{t-1} + \phi_{3,2} \Delta u_{t-2} + \phi_{3,3} \Delta u_{t-3} \\ & + \phi_{4,1} \Delta(p_O - p_T)_{t-1}) G(CV_{t-1}; \gamma, c_1, c_2) \end{aligned}$$

$$\text{where } G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-\gamma (CV_{t-1} - c_1) (CV_{t-1} - c_2) / \sigma^2(CV_{t-1})]\}^{-1},$$

is the ‘quadratic logistic’ transition function, with  $CV_{t-1}$  as the transition variable. Values of 0 and 1 of the transition function are associated with the two alternative regimes. The  $\Delta(w - p_T)_t$  dynamics in the first regime, when  $G(CV_{t-1}; \gamma, c_1, c_2) = 0$ , are:

$$\Delta(w - p_T)_t = (\mu_1 + \alpha_1 CV_{t-1} + \phi_{1,1} \Delta(p_T - p)_{t-1} + \phi_{2,1} \Delta(p_O - p_T)_{t-1}) (1 - G(CV_{t-1}; \gamma, c_1, c_2)).$$

In the second regime, when  $G(CV_{t-1}; \gamma, c_1, c_2) = 1$ , its dynamics are:

$$\begin{aligned} \Delta(w - p_T)_t = & (\mu_2 + \phi_{3,1} \Delta(p_T - p)_{t-1} + \phi_{3,2} \Delta u_{t-2} + \phi_{3,3} \Delta u_{t-3} \\ & + \phi_{4,1} \Delta(p_O - p_T)_{t-1}) G(CV_{t-1}; \gamma, c_1, c_2). \end{aligned}$$

For intermediate values of  $G(CV_{t-1}; \gamma, c_1, c_2)$ , i.e.  $0 < G(CV_{t-1}; \gamma, c_1, c_2) < 1$ ,  $\Delta(w - p_T)_t$  dynamics are a weighted average of the two equations. The speed of transition between the two regimes is determined by the parameter  $\gamma$ .

$$\begin{aligned} \Delta(w - p_T)_t = & \begin{pmatrix} 0.008 & +0.102 CV_{t-1} & -0.343 \Delta(p_T - p)_{t-1} \\ (0.002) & (0.062) & (0.083) \end{pmatrix} \\ & -0.331 \Delta(p_O - p_T)_{t-1} (1 - G(CV_{t-1}; \gamma, c_1, c_2)) \\ & (0.098) \\ & \begin{pmatrix} +0.015 & -1.600 \Delta(p_T - p)_{t-1} & -0.933 \Delta u_{t-2} \\ (0.011) & (0.306) & (0.248) \end{pmatrix} \\ & +1.040 \Delta u_{t-3} \quad -0.656 \Delta(p_O - p_T)_{t-1} G(CV_{t-1}; \gamma, c_1, c_2) \\ & (0.200) \quad (0.298) \end{aligned}$$

where

$$G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \begin{pmatrix} \exp[-2.829(CV_{t-1} & +0.070) (CV_{t-1} & -0.103) / \sigma^2(CV_{t-1})] \\ (1.657) & (0.007) & (0.014) \end{pmatrix}\}^{-1}$$

$$s_{NL} = 0.021, \quad s_{NL}^2 / s_L^2 = 0.705, \quad \text{AR}(5) = 0.84[0.525], \quad \text{ARCH}(4) = 0.64[0.635], \quad \text{HET} = 0.71[0.813],$$

$$\text{NORM}(2) = 13.57[0.001]$$

Notes: Standard errors are given in parentheses below the estimates.  $s_{NL}$ : standard error of the non-linear regression. The diagnostic tests are discussed in the notes of Table 3.

Table 6

Estimated non-linear  $\Delta u_t$  model

The Table reports the NLS estimates of the following STVECM equation:

$$\begin{aligned} \Delta u_t = & (\mu_1 + \phi_{1,1}\Delta(p_T - p)_{t-1} + \phi_{1,2}\Delta(w - p_T)_{t-1} + \phi_{1,3}\Delta u_{t-1} \\ & + \phi_{2,1}\Delta(p_O - p_T)_{t-1})(1 - G(CV_{t-1}; \gamma, c_1, c_2)) \\ & + (\mu_2 + \alpha_2 CV_{t-1} + \phi_{3,1}\Delta u_{t-1} + \phi_{3,2}\Delta u_{t-3} + \phi_{4,1}\Delta(p_O - p_T)_t)G(CV_{t-1}; \gamma, c_1, c_2) \end{aligned}$$

where  $G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-\gamma (CV_{t-1} - c_1) (CV_{t-1} - c_2) / \sigma^2(CV_{t-1})]\}^{-1}$ ,

is the ‘quadratic logistic’ transition function, with  $CV_{t-1}$  as the transition variable. Values of 0 and 1 of the transition function are associated with the two alternative regimes. The  $\Delta u_t$  dynamics in the first regime, when  $G(CV_{t-1}; \gamma, c_1, c_2) = 0$ , are:

$$\begin{aligned} \Delta u_t = & (\mu_1 + \phi_{1,1}\Delta(p_T - p)_{t-1} + \phi_{1,2}\Delta(w - p_T)_{t-1} + \phi_{1,3}\Delta u_{t-1} \\ & + \phi_{2,1}\Delta(p_O - p_T)_{t-1})(1 - G(CV_{t-1}; \gamma, c_1, c_2)) \end{aligned}$$

In the second regime, when  $G(CV_{t-1}; \gamma, c_1, c_2) = 1$ , its dynamics are:

$$\Delta u_t = (\mu_2 + \alpha_2 CV_{t-1} + \phi_{3,1}\Delta u_{t-1} + \phi_{3,2}\Delta u_{t-3} + \phi_{4,1}\Delta(p_O - p_T)_t)G(CV_{t-1}; \gamma, c_1, c_2).$$

For intermediate values of  $G(CV_{t-1}; \gamma, c_1, c_2)$ , i.e.  $0 < G(CV_{t-1}; \gamma, c_1, c_2) < 1$ ,  $\Delta u_t$  dynamics are a weighted average of the two equations. The speed of transition between the two regimes is determined by the parameter  $\gamma$ .

$$\begin{aligned} \Delta u_t = & \begin{pmatrix} 0.001 & -0.394 \Delta(p_T - p)_{t-1} & -0.443 \Delta(w - p_T)_{t-1} & +0.814 \Delta u_{t-1} \\ (0.003) & (0.268) & (0.246) & (0.071) \end{pmatrix} \\ & -0.174 \Delta(p_O - p_T)_{t-1} (1 - G(CV_{t-1}; \gamma, c_1, c_2)) \\ & (0.127) \\ & \begin{pmatrix} +0.009 & -0.291 CV_{t-1} & +0.991 \Delta u_{t-1} & -0.473 \Delta u_{t-3} \\ (0.004) & (0.063) & (0.087) & (0.089) \end{pmatrix} \\ & -0.384 \Delta(p_O - p_T)_t G(CV_{t-1}; \gamma, c_1, c_2) \\ & (0.169) \end{aligned}$$

where

$$G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-11.001(CV_{t-1} + 0.040)(CV_{t-1} - 0.052) / \sigma^2(CV_{t-1})]\}^{-1}$$

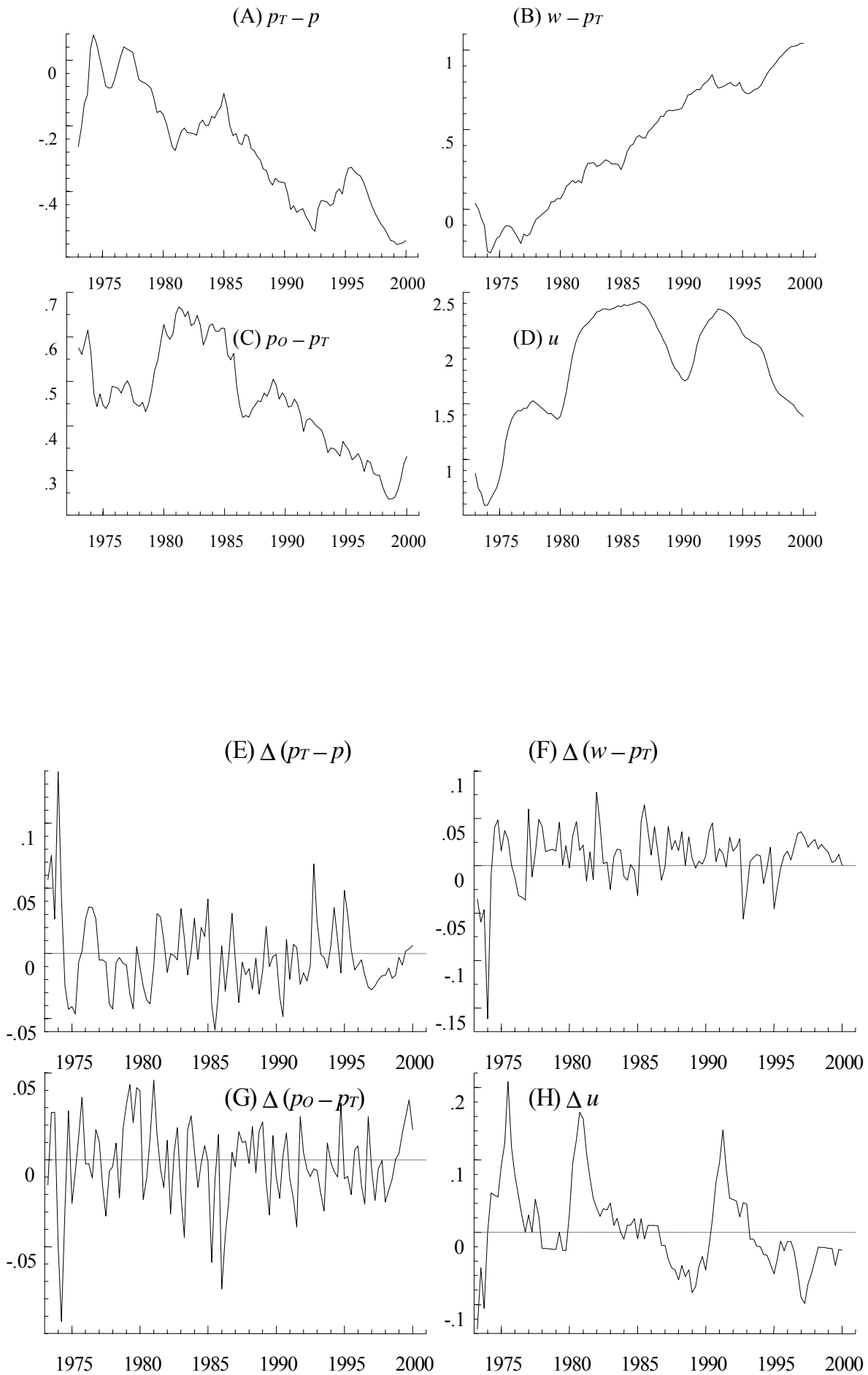
(8.196)                      (0.003)                      (0.008)

$$s_{NL} = 0.022, \quad s_{NL}^2/s_L^2 = 0.840, \quad AR(5) = 1.43[0.223], \quad ARCH(4) = 3.26[0.020], \quad HET = 1.30[0.200],$$

$$NORM(2) = 12.78[0.002]$$

Notes: Standard errors are given in parentheses below the estimates.  $s_{NL}$ : standard error of the non-linear regression. The diagnostic tests are discussed in the notes of Table 3.

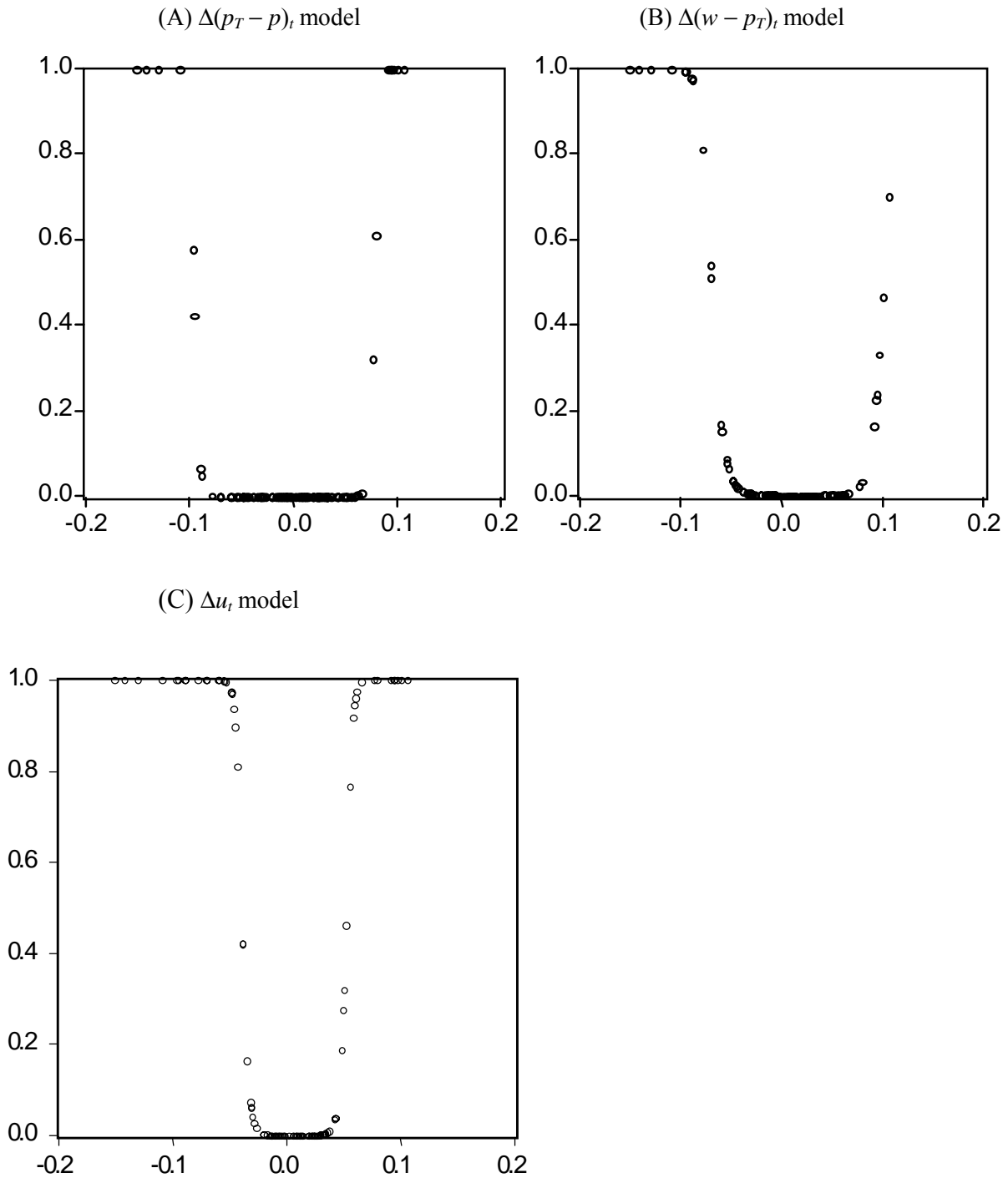
**Figure 1:** Plots of the levels and the first differences of the series



**Figure 2:** Deviations from the estimated long-run real exchange rate relationship



**Figure 3:** Estimated transition functions (vertical axis) against  $CV_{t-1}$  (horizontal axis)



Notes: Panels (A), (B) and (C) report the estimated transition functions for  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  against the transition variable  $CV_{t-1}$  from the corresponding STVECM equations as reported in Tables 4 to 6. The estimated transition functions are:

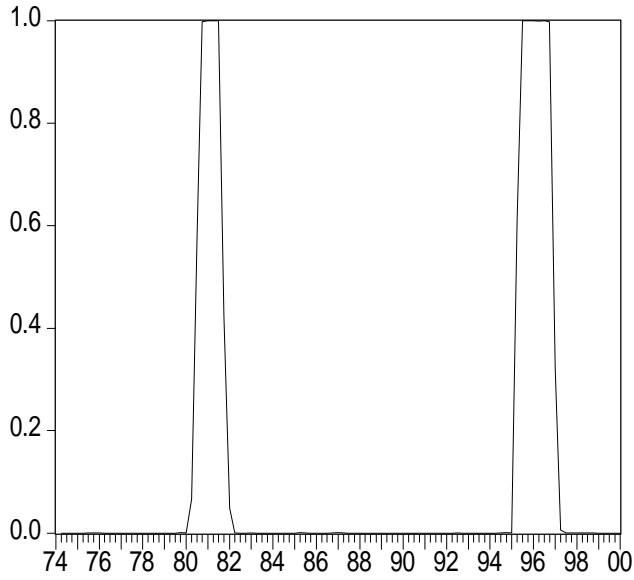
$$(A) G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-7.016(CV_{t-1} + 0.096) (CV_{t-1} - 0.078)/ \sigma^2(CV_{t-1})]\}^{-1}$$

$$(B) G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-2.829(CV_{t-1} + 0.070) (CV_{t-1} - 0.103)/ \sigma^2(CV_{t-1})]\}^{-1}$$

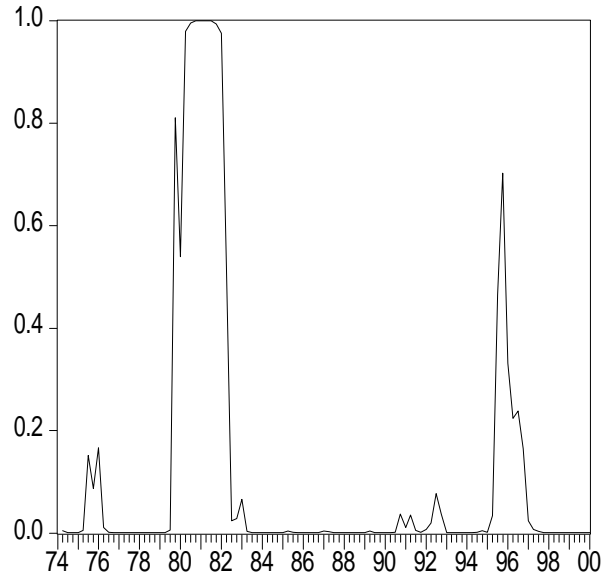
$$(C) G(CV_{t-1}; \gamma, c_1, c_2) = \{1 + \exp[-11.001(CV_{t-1} + 0.040) (CV_{t-1} - 0.052)/ \sigma^2(CV_{t-1})]\}^{-1}$$

**Figure 4:** Estimated transition functions against time

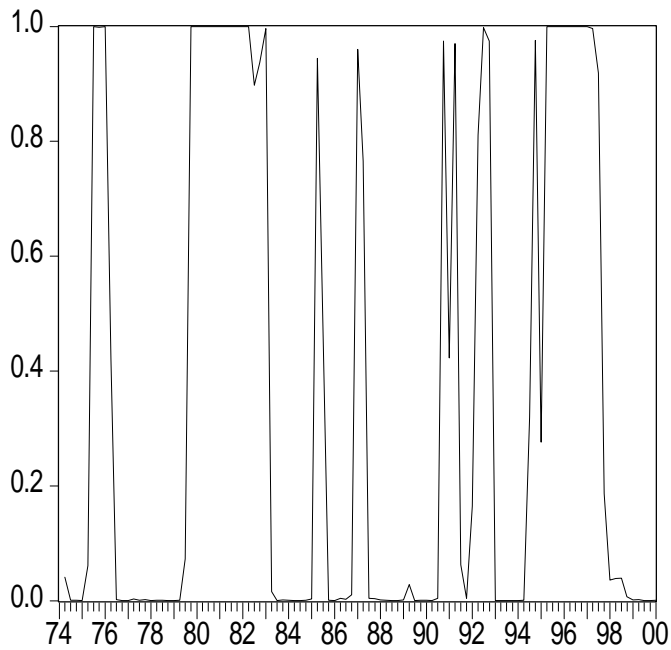
(A)  $\Delta(p_T - p)_t$  model



(B)  $\Delta(w - p_T)_t$  model



(C)  $\Delta u_t$  model



Notes: Panels (A), (B) and (C) report the estimated transition functions for  $\Delta(p_T - p)_t$ ,  $\Delta(w - p_T)_t$  and  $\Delta u_t$  against time from the corresponding STVECM equations as reported in Tables 4 to 6 and Figure 3. Extreme values of 0 and 1 of the transition functions are associated with the two alternative regimes.