Purchasing Power Parity between the UK and the Euro Area

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Purchasing Power Parity between the UK and the Euro Area

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Abstract: We use the Johansen cointegration approach to assess the empirical validity of the purchasing power parity (PPP) between the UK and the Euro Area, which we represent by Germany, the largest of its members. We conduct the empirical analysis in the context of the global financial crisis that began in 2007 and find that it directly affects the cointegration space. We fail to validate the Johansen and Juselius (1992) original hypothesis that nonstationarity of the PPP associates with the nonstationarity of interest rate differentials to produce a stationary relation. On the other hand, we do not reject PPP. We find that PPP cointegrates with inflation differentials. We also find, contrary to conventional wisdom, that (i) equilibrium adjustment occurs between the German and UK inflation rates, while weak exogeneity exists for the German and UK interest rates and the PPP condition, and (ii) three common trends associated with the German interest rate, the UK interest rate, and the PPP condition “push” the system with the German interest rate and the PPP condition playing dominant roles.

Keywords: Purchasing Power Parity, Euro Area, Cointegrated VAR.
JEL Classifications: E31, E43, F31, F32

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1. Introduction

Purchasing power parity (PPP) proves a most controversial hypothesis in the international finance literature. A key component of a number of theoretical models, such as flexible-price monetary models (Frankel, 1976; Mussa, 1976), sticky-price monetary models (Dornbusch, 1976), international asset pricing models (Merton, 1973; Solnik, 1974), and new open economy macroeconomic models (Obstfeld and Rogoff, 1995), PPP postulates that exchange rates adjust in the long run to price differentials in open economies to restore international commodity market equilibrium. A relatively extensive literature exists that examines the empirical validity of the PPP condition, mainly over the post-Bretton Woods system of floating nominal exchange rates. The empirical evidence, however, is rather mixed, and varies depending upon the time period, the countries and the econometric methodology. For a detailed overview see, for example, the recent survey studies by Rogoff (1996), Sarno and Taylor (2002), and Taylor and Taylor (2004).

Most empirical work centers on tests for a unit root in the real exchange rate or for cointegration between domestic and foreign prices and the nominal exchange rate. For instance, early studies, such as Adler and Lehmann (1983), Huizinga (1987), Edison (1987), and Corbae and Ouliaris (1988) find that the real exchange rate does not exhibit a stationary process. In contrast, Kim (1990), Glen (1992), Grilli and Kaminsky (1991), Lothian and Taylor (1996), Kuo and Mikkola (1999), and Chen and Wu (2000), using long time periods, find that real exchange rate reverts to its mean (although with a high degree of persistence) and follows a stationary process. The search for cointegration between the nominal exchange rate and domestic and foreign prices, although more limited, also provides conflicting evidence (e.g., Taylor, 1988; Mark, 1990; Layton and Stark, 1990; Cheung and Lai, 1993; Ender and Falk, 1998; and Coakley
and Fuertes, 2000). Recent advances in panel econometrics (e.g., Papell, 1997; Frankel and Rose, 1996; Pedroni, 1995, 2001; Lothian, 1997; MacDonald, 1996; Wu and Chen, 1999; and Taylor and Sarno, 1998), nonlinear dynamics (e.g., Holmes and Maghrebi, 2004; Baum, Barkoulas, and Caglayan, 2001; and Kilian and Taylor, 2003), and cross-sectional dependence (e.g., O’Connell, 1998; Harris, Leybourne, and McCabe, 2005; Chortareas and Kapetanios, 2009; and Snaith, 2012) find stronger evidence supporting PPP, though the empirical findings remain mixed.

In a path-breaking paper, Johansen and Juselius (1992) attribute the apparent failure of PPP to the lack of precise specification of the sampling distribution of the data. That is, the research generally neglects (i) the time-series properties of the data, (ii) the possible interactions between prices, interest rates, and exchange rates, and (iii) differences between short-run and long-run effects. In particular, using the methodology of multivariate cointegration developed by Johansen (1988, 1991), Johansen and Juselius (1992) find support for PPP using data for the UK, but only when uncovered interest rate parity (UIP) appears in the system. Since the seminal paper of Johansen and Juselius (1992), a growing recognition emerges that while PPP does not hold in isolation, a long-run stationary relationship can occur between the real exchange rate and interest rate differentials. Juselius (1995) uses the same framework to analyze the mechanisms explaining the inflationary effects transmitted from Germany to Denmark and finds that the link between the goods and asset markets, postulated by the combined relation of PPP and UIP, is crucial for a full understanding of the movements of exchange rates, prices, and interest rates. Juselius and MacDonald (2000, 2004) apply this methodology to investigate the international parity relationships between the US and Germany, and the US and Japan. They argue that the balance of payment constraint implies that the financing of any imbalance in the current account must come from the capital and financial account. Hunter (1992), Sjoo (1995), Pesaran, Shin,
and Smith (2000), Miyakoshi (2004), Ozmen and Gokcan (2004), Caporale, Kalyvitis, and Pitts (2001), Camarero and Tamarit (1996), Hatzinikolaou and Polasek (2005), among others, provide further evidence that modeling the interactions between the PPP and UIP generates linear stationary relations.

We extend this burgeoning literature by examining the empirical validity of PPP through the interdependence of adjustments in the international asset and commodity markets using data from the UK and the Euro Area. More precisely, we investigate whether the PPP holds for the exchange rate between the UK (“foreign country”) and Germany (“home country”) when combined with UIP. We represent the Euro Area by Germany, since Germany is the largest economy within the European Union and the Euro Area, as well as one of the major trading partners of the UK.¹ We employ the Johansen cointegration method, which provides a flexible class of statistical models that combine long-run cointegrating relationships and short-run dynamics. We use monthly data for the UK and Germany, spanning the period from the introduction of the euro in January 1999 through April 2011. Using this sample period includes the recent period of the global financial crisis originating from the collapse of the US housing market in 2007, we explore the stability of the “augmented” system where inflation differentials and interest rate differentials enter the long-run cointegrating relationships.

The econometric analysis receives motivation from at least two pragmatic considerations. First, Germany and the UK are members of the European Union, a custom union and a common market that eliminated most trade barriers and capital controls among its members. This, in turn, virtually removes a large number of impediments that can prevent PPP and UIP from holding.

¹ Germany and the UK form a most important trade relationship. In 2011, only the US surpassed Germany in UK exports and Germany did hold the position as the top trading partner for imports, accounting for 11.06% and 12.87% of the UK’s primary exports and imports, respectively. In comparison, the US accounted for 14.71% and 9.74%, respectively.
Thus, intuitively, we expect PPP to hold the best between Germany and the UK. Historically, however, the UK resisted a deep involvement with the European Economic Community (EEC). The 1957 Treaty of Rome established the EEC, but the UK only joined in 1973. The European Monetary System (EMS), founded in 1979, created the Exchange Rate Mechanism (ERM), a fixed exchange rate arrangement designed to reduce exchange rate volatility and achieve monetary stability in preparation for the introduction of the euro. The UK left the ERM in 1992, showing its economic independence, freeing the pound from the ERM, and regaining control over its monetary policy and interest rates. Since then, the UK resisted rejoining any type of exchange rate regime with the Euro Area countries and joining in the adoption of the euro.

Second, an obvious and intense interest exists within Europe as to the nature of the links between the countries of the European Union. The recent literature on the “European” business cycle (e.g., Barrios, Brülhart, Elliott, and Sensier, 2003; Kontolemis and Samiei, 2000; Camacho, Perez-Quiros, and Saiz, 2008) suggests that the UK and the Euro Area do not exhibit converging and synchronous business cycles. Barrios, Brülhart, Elliott, and Sensier (2003) find that the UK business cycle remains persistently out of phase with that of the main Euro Area economies. Kontolemis and Samiei (2000) provide evidence that the UK business cycle achieves its relative independence from the Euro Area economies because of its independent monetary policy. Camacho, Perez-Quiros, and Saiz (2008) find that UK business cycles more closely match the business cycles of Canada and US than the business cycles of the Euro area countries. Moreover, they show that no evidence exists of a “European economy” that acts as an attractor to the other economies of the area.

Accordingly, in light of the historical developments of the economic and financial relations between the UK and the European Union, and the findings of diverging and asynchronous business cycles, the question of the current degree of economic integration
between the UK and the Euro Area becomes important.

Empirical analysis and tests of PPP and UIP shed some light on this issue. The PPP and UIP conditions provide indicators of the degree of economic integration between economies. PPP measures integration of the commodity markets, whilst UIP measures financial integration, and the greater the economic integration across countries, the greater the likelihood that these conditions will receive empirical support.

Few empirical studies (Alquist and Chinn, 2002; Gadea, Montañés, and Reyes, 2004; Lopez and Papell, 2007) examine PPP within the Euro Area using “synthetic” euro data.² Alquist and Chinn (2002) find that the real exchange rate is nonstationary, suggesting that PPP does not hold in the Euro Area. Gadea, Montañés, and Reyes (2004) find some support for PPP within the Euro Area after incorporating two structural breaks. Scant evidence of the validity of PPP between the Euro Area and other major economies exists. Lopez and Papell (2007) study the convergence to PPP in the Euro Area from 1973 to 2001 and find that PPP holds better within the Euro Area than between the Euro Area and other European countries. Chinn (2002), using data on the “synthetic” euro-dollar exchange rate for 1985 to 2001, rejects PPP, but documents a stable long-run relationship between the real euro-dollar rate, productivity differentials, and the real price of oil. Koedijk, Tims, and van Dijk (2004), in addition to examining the validity of PPP within the Euro Area, also use “synthetic” euro data to study the validity of PPP between the Euro Area and other major economies. They find that, with the exception of Switzerland, PPP does not hold. Manzur and Chan (2010), using data through April 2007, construct a measure of “pooled” inflation among the 12 Euro countries and use this measure to test, in a simple regression framework, the relative version of PPP for the euro against the currencies of Japan,

² The “synthetic” euro is an artificial exchange rate constructed as a geometrically weighted average of the exchange rates of individual EMU currencies prior to 1999 (Artis and Beyer, 2004).
the UK, and the US. Their results provide weak support for PPP in the case of USD/Euro and £/Euro exchange rates, and rejects PPP for the Yen/Euro.

We organize the rest of this paper as follows. Section 2 discusses the economic model and the statistical restrictions implied by the PPP and UIP conditions. Section 3, after a brief description of the data, performs a comprehensive I(1) cointegrated vector autoregressive (VAR) analysis, discusses the long-run effects of the stochastic trends, and conducts a long-run impact analysis. Section 4 offers concluding remarks.

2. The Economic Framework

According to the PPP condition, the nominal exchange rate between currencies of two countries depends on the relative prices in home and foreign countries. In its simplest form, absolute PPP is defined as:

\[ S_t = \frac{P_t}{P_t^*} \]  

(1)

and deviations from PPP with continuous compounding are defined as:

\[ \text{ppp}_t = p_t - p_t^* - s_t, \]  

(2)

where \( p_t \) is the logarithm of the domestic price level \( P_t \), \( p_t^* \) is the logarithm of the foreign price level \( P_t^* \), and \( s_t \) denotes the logarithm of the exchange rate \( S_t \) (measured as units of domestic currency per unit of foreign currency ). In empirical applications, we verify PPP if \( \text{ppp}_t \) is stationary.

According to the UIP condition, the interest rate differential between two countries equals the expected change in the exchange rate. In its simplest form, the UIP condition is defined as follows:

\[ E_t(\Delta s_{t+1}) = i_t - i_t^*, \]  

(3)
where $\Delta$ is the first-difference operator, $E_t$ denotes the conditional expectation operator at time $t$ based on information available at time $t-1$, $E_t(\Delta s_{t+1})$ equals the expected depreciation rate of the nominal exchange rate from period $t$ to $t+1$, $i_t$ is the domestic interest rate, and $i_t^*$ is the foreign interest rate. Logarithmic differencing equation (1) and applying the expectation operator gives:

$$E_t(\Delta s_{t+1}) = E_t(\Delta p_{t+1}) - E_t(\Delta p^*_{t+1}), \quad (4)$$

which gives relative PPP.

Substituting equation (4) into equation (3) and rearranging terms yields:

$$i_t - i_t^* - E_t(\Delta p_{t+1}) + E_t(\Delta p^*_{t+1}) = 0. \quad (5)$$

Under the assumption of rational expectations, where agents do not make systematic forecast errors in inflation rates, then

$$E_t(\Delta p_{t+1} - \Delta p^*_{t+1}) = \Delta p_t - \Delta p_t^* + \nu_t, \quad (6)$$

where $\nu_t$ denotes an unpredictable i.i.d. shock. Then, combining equations (5) and (6) leads to

$$i_t - i_t^* - \Delta p_t + \Delta p_t^* = \nu_t \quad (7)$$

and testing for the UIP condition amounts to testing whether $\nu_t$ is stationary.

Following Juselius (1995) and Juselius and MacDonald (2000, 2004), we write a relation that combines the PPP and the UIP as follows:

$$(i_t - i_t^*) = (\Delta p_t - \Delta p_t^*) + ppp_t + \nu_t. \quad (8)$$

It follows that the PPP and UIP conditions hold jointly if $(i_t - i_t^*) - (\Delta p_t - \Delta p_t^*) - ppp_t$ is stationary. This can occur either if jointly $i_t - i_t^* \sim I(0)$, $\Delta p_t - \Delta p_t^* \sim I(0)$, and $ppp_t \sim I(0)$ or if $i_t - i_t^* \sim I(1)$, $\Delta p_t - \Delta p_t^* \sim I(1)$, and $ppp_t \sim I(1)$, but their linear combination
\((i_t - i^*_t) - (\Delta p_t - \Delta p^*_t) - ppp_t \sim I(0)\). In the first case, the PPP and UIP conditions hold independently of each other; in the second case, instead, they do not hold individually, but do hold together. That is, the nonstationarity of the PPP condition associates with the nonstationarity of the UIP condition to produce a stationary relation. We can also interpret equation (8) as follows: the nonstationarity of inflation differentials and interest rate differentials removes the nonstationarity of \(ppp_t\) (i.e., the movements in inflation differentials, interest rate differentials, or both, compensate deviations from PPP).

Equation (8) defines a stationary equilibrium relation where interest rates and inflation rates pull the system together whenever the economy pushes \(ppp_t\) away from equilibrium. A more flexible formulation, which relaxes the rational expectations hypothesis and acknowledges the weak correspondence between theoretical and observed variables and the effect of temporal aggregation (Juselius, 1995), leads to equation (9):

\[
\omega_1(i_t - i^*_t) - \omega_2(\Delta p_t - \Delta p^*_t) - \omega_3 ppp_t \sim I(0),
\]

where \(\omega_1, \omega_2,\) and \(\omega_3\) are weights on PPP and UIP, which depend on the underlying structural parameters. The stationarity of the PPP and UIP conditions emerge as a special case of equation (9) when we set \(\omega_1\) and \(\omega_2\) equal to zero and set \(\omega_3\) equal to one, or we set \(\omega_3\) equal to zero and set \(\omega_1\) and \(\omega_2\) equal to one, respectively. Other special cases of equation (9) define stationary equilibrium relations, where either (i) \(\omega_1(i_t - i^*_t) - \omega_3 ppp_t \sim I(0)\) and \(\omega_2 = 0\) or (ii) \(-\omega_2(\Delta p_t - \Delta p^*_t) - \omega_3 ppp_t \sim I(0)\) and \(\omega_1 = 0\). In case (i), interest rates pull the system whenever the economy pushes \(ppp_t\) away from equilibrium; while in case (ii), inflation rates pull the system together whenever the economy pushes \(ppp_t\) away from equilibrium.
Juselius (1995) and Juselius and MacDonald (2000, 2004) emphasize case (i). They argue that models of exchange rate determination pertaining to these economies must jointly consider the deviations from PPP and UIP parities to induce stationarity by including the interaction of goods and capital markets. Pedersen (2002a, 2002b), on the other hand, proposes case (ii), which he calls “PPP with adjustment”. That is, he argues, based on Gregory, et al. (1993) and Gregory (1994) that deviations from $\text{ppp}_i$ will adjust back to equilibrium, but not without adjustment costs. Specifically, PPP with adjustment holds, which Pederson (2002b) writes as “Definition 3”, when two I(2) price levels exhibit the following relationship (in our notation):

$$\text{ppp}_i + k_1 \Delta p_i + k_2 \Delta p_i^* \sim I(0)$$

and $k_1 \neq 0, k_2 \neq 0, k_1 \neq k_2$. This relationship assumes asymmetric adjustment costs, and can be extended to include the case of symmetric adjustment costs:

$$\text{ppp}_i + k(\Delta p_i - \Delta p_i^*) \sim I(0),$$

$\forall k \neq 0$. In such a case, the inflation rate differential represents the adjustment costs. Deviations from $\text{ppp}_i$ indicate the degree of market integration. For perfect integration, $\text{ppp}_i$ equals one. Less than perfect integration leaves $\text{ppp}_i$ different from one. If so, then the cost of adjusting back to perfect integration depends on the inflation rate differential.

3. **The Empirical Analysis**

3.1 **The data and their univariate properties**

The empirical analysis uses monthly data for Germany and the UK over the period January 1999 to April 2011 (148 observations). The variables used in the analysis are defined as follows: $\Delta p_i$ = the German inflation rate, $\Delta p_i^*$ = the UK inflation rate, $i_i$ = the German 10-year constant maturity bond yield, $i_i^*$ = the UK 10-year constant maturity bond yield, $\text{ppp}_i = p_i - p_i^* - s_i$ where $p_i$ = the log of the German price index, $p_i^*$ = the log of the UK price index, $s_i$ = the log of the nominal exchange rate, defined as Euro/£. The price indices are the harmonized indices of consumer
prices (HCPI), 2005=100, and are not seasonally adjusted. We compute the rates of inflation as the logarithmic first difference of consumer prices. Data on exchange rate and price indices come from the statistical database of the European Central Bank (sdw.ecb.europa.eu), while data on bond yields come from the OECD Main Economic Indicators database (stats.oecd.org). We convert annual interest rates to monthly rates and divide by 100 to make the estimates comparable with logarithmic monthly changes. For similar reasons, we divide the ppp, series by 100. Since ppp, is the negative of the natural logarithm of the real exchange rate, a positive trend in ppp, means a real appreciation of the euro, although a rise in the exchange rate series means a nominal depreciation of the euro.

The Johansen cointegration procedure requires nonstationary variables. We ascertain the order of integration of each of the five series and their first differences using the standard Augmented Dickey-Fuller (ADF) and the more efficient Dickey-Fuller-Generalized Least Squares (DF-GLS) tests. Dickey and Fuller (1979) and Elliott, Rothenberg, and Stock (1996) present the details of these tests. The DF-GLS test procedure applies the DF test to locally demeaned (or demeaned and detrended) series and generally exhibits higher power than the standard ADF unit-root test. We choose the lag length of the ADF and DF-GLS tests using the Akaike information criterion (AIC) with an upper bound of 13 lags. We include a constant but no deterministic time trends. The inclusion of a linear trend proves insignificant and does not modify the main results in any substantial manner. The results of both the ADF and DF-GLS tests, reported in Tables 1 and 2, do not reject nonstationarity for all series using the 5-percent level. Strong evidence exists that all series possess a unit root or are I(1) processes (i.e.,
nonstationary in levels, but stationary in first differences). Many other studies contain the finding of a unit root in inflation rates (e.g., Rapach, 2003; Banerjee, Cockerell, and Russel, 2001). Based upon these findings, we test for unit roots in inflation differentials \( (\Delta p_t - \Delta p_t^*) \), interest rate differentials \( (i_t - i_t^*) \), German \( (i_t - \Delta p_t) \) and UK \( (i_t^* - \Delta p_t^*) \) real interest rates, and real interest rate spreads, \( (i_t - \Delta p_t) - (i_t^* - \Delta p_t^*) \). Theory suggests stationarity of interest rate differentials (i.e., interest rate parity) and stationarity of real interest rate spreads (i.e., real interest rate parity). Tables 1 and 2 report that we cannot reject the null hypotheses of unit roots in these series using a 5-percent level.

Our empirical approach starts from a statistically well-specified five-dimensional unrestricted VAR model for the components of \( x_t = [\Delta p_t, \Delta p_t^*, i_t, i_t^*, ppp_t] \sim I(1) \), and then reduces this general statistical model by testing for various theoretical restrictions. That is, our modeling approach responds to the economic questions of interest by embedding the economic model within the statistical model and using strict statistical principles as criteria to determine the adequacy of various empirical models.

In any VAR framework, the chosen lag length can importantly affect the results, since all inferences in both the cointegration and common trends analysis depend on the number of lags specified. No definitive procedure exists for choosing the lag length. Standard criteria defined by the multivariate versions of the Akaike information criterion (AIC), the Bayesian information criterion (BIC), and the Hannan-Quinn (H-Q) criterion suggest a lag length of 2, given a maximum lag order of 4. We cannot justify the VAR(2) specification, however, as diagnostic

\[ \frac{\text{Levin, Lin, and Chu test statistic}}{\text{Breitung test statistic}} = \frac{0.796}{-1.046} \]

\[ \frac{\text{Fisher Chi-square (ADF) statistic}}{\text{Choi Z-statistic (ADF)}} = \frac{9.814}{-0.294} \]

Since the power of univariate unit-root tests is notoriously low, we also conducted a series of panel unit-root tests. The Levin, Lin, and Chu test statistic equals 0.796 (\( p\)-value = 0.787), the Breitung test statistic equals -1.046 (\( p\)-value = 0.147), and the Fisher Chi-square (ADF) statistic equals 9.814 (\( p\)-value = 0.456), while the Choi Z-statistic (ADF) equals -0.294 (\( p\)-value = 0.384). The first two tests assume a common unit-root process, while the last two tests assume individual unit-root processes.
tests suggest residual serial correlation. Consequently, we specify a VAR(3) model, using a number of specification tests.4

A further issue concerns the appropriate treatment of deterministic components such as constant and trend term (i.e., whether deterministic variables should enter the cointegrating space or the short-run model). Different treatment of constant and trend terms in the analysis lead towards different critical values (Johansen, 1991; Johansen and Juselius, 1990). We specify the model to include a restricted constant, following Johansen’s (1995) suggestion that if the variables included in the system do not show growth, then the constant term should appear in the cointegrating space, implying that some equilibrium means in the cointegration space can differ from zero. We do not include a linear deterministic trend, since a trend is inconsistent with PPP (Papell and Theodoridis, 1998; Amara and Papell, 2006). Excluding a linear deterministic trend also proves consistent with the unit-root analysis.

We include three different types of dummy variables. First, we introduce centered seasonal dummy variables $D_t$ to account for seasonality in the data. Johansen (1996) proposes centered dummy variables, since they sum to zero over a year and are orthogonalized on the constant term. Second, we use a shift dummy variable $C_{2007:10}$ to account for the developments of the global economic crisis.5 That is, $C_{2007:10}$ equals 0 before October 2007 and 1 from October 2007 onward. We restrict this dummy variable to lie in the cointegration space to allow for the possibility that the global financial crisis of 2007 exerts a permanent effect on the mean of the cointegrating relations over the sample period. This issue is particularly relevant to our analysis

4 This implies 2 lags of the first differences of the variables in the VEC model of the data.

5 We assume that the break occurs in 2007:10 based on the visual inspection of the graphs combined with the institutional consideration about the beginning of the global recession and sub-prime crisis in the housing markets. We cannot verify, however, that the break really occurs at that time.
since Stephon and Larsen (1991) show that the cointegration tests may reflect sample dependency. Finally, we include an intervention dummy variable $D_{2008:12}$ to account for a residual exceeding in absolute value $3\sigma_e$.

The final specification of the unrestricted VAR(3) model in error correction form is as follows:

$$
\Delta x_t = \Gamma_1 \Delta x_{t-1} + \Gamma_2 \Delta x_{t-2} + \alpha(\beta' x_{t-1} + \rho_0 + \rho_1 C_{2007:8}) + \phi_0 D_{2008:12} + \phi_1 D_t + \varepsilon_t, \quad (10)
$$

$$
\varepsilon_t \sim N_p(0, \Sigma), \ t = 1, \ldots, T
$$

where $(\Gamma_1, \Gamma_2, \alpha, \beta, \rho_0, \rho_1, \Sigma)$ are unrestricted, $\Delta x_t' = [\Delta^2 p_t, \Delta^2 p_t^*, \Delta i_t, \Delta i_t^*, \Delta ppp_t]$, $\rho_0$ is the constant restricted to the cointegration space, and $\rho_1$ is the coefficient of the shift dummy variable. The model in equation (10) is greatly overparameterized, but represents the starting point from which we test the various structural hypotheses (such as rank restrictions and linear parameter restrictions) extracted from economic theory. We perform cointegration analysis and related calculations and graphs using CATS in RATS, version 2 detailed in Dennis, Hansen, Johansen, and Juselius (2006).

### 3.2 Specification tests

This section summarizes the results from a battery of diagnostic tests recommended in Johansen and Juselius (1990, 1992) and Juselius (2006) applied to the residuals of equation (10). Although the estimated coefficients of Equation (10) do not necessarily conform to economic interpretations, the unrestricted VAR provides the starting point from which a more parsimonious and economically meaningful representation can emerge. We must adequately specify the residual structure of the unrestricted VAR(3), however, prior to the determination of the cointegration rank and prior to the imposition of coefficient restrictions that emerge from a
series of formal hypothesis tests. We make this assessment using the array of multivariate and univariate tests for serial correlation, autoregressive conditional heteroskedasticity (ARCH), and normality reported in Table 3.

The results of the multivariate Lagrange Multiplier (LM) test (Hosking, 1980) as implemented in Johansen (1995) indicate that we cannot reject the null hypotheses of no first- and second-order autocorrelation at the 1- and 5-percent significance levels, respectively.

The multivariate and univariate normality tests (Doornik and Hansen, 2008; Ljung and Box, 1978) look for skewness and kurtosis. We cannot reject the hypothesis of normality both in the multivariate case and the univariate case at any conventional level in all cases, except the univariate test of \( \text{ppp}_1 \), where we cannot reject at the 5-percent level. This is an important finding, since the properties of the cointegration estimators are sensitive to deviations from normality, especially deviations due to skewness.

The evidence from the multivariate ARCH tests (Duchesne and Lalancette, 2003; Hacker and Abdulnasser, 2005) suggests some problems with conditional heteroskedasticity. The univariate tests (Engle, 1988), however, show no signs of ARCH effects at the 10-percent level. Table 3 also reports the \( R^2 \) statistic, which measures the improvement in the explanatory power of the model compared to the random walk hypothesis. The model better explains changes in inflation rates than changes in interest rates and \( \text{ppp}_1 \).

Figures 1 to 5 plot the actual and fitted values, the standardized residuals, the histogram of the standardized residuals with a superimposed histogram of the standardized normal distribution, and the correlogram for lags from 1 to 36. The graphical analysis suggests that the model exhibits well-behaved standardized residuals. We, therefore, conclude that the unrestricted VAR(3) model with a restricted constant, a shift dummy variable, an intervention dummy
variable, and centered seasonal dummy variables provides satisfactory diagnostic results as a whole. Thus, we use this model for the I(1) cointegration analysis.

3.3 Determination of the cointegrating rank: the trace test

Johansen (1988) shows that the existence of cointegrating vectors implies that the system exhibits reduced rank. The cointegrating rank divides the data into \( r \) linearly independent cointegrating relationships and \( p - r \) common stochastic trends. We can interpret the cointegrating relationships as pulling the system through an adjustment process to long-run equilibrium. We can interpret the common trends, on the other hand, as the components of the system that push the process. Consequently, the cointegration rank proves crucial to the analysis, and affects all remaining inferences. Since distinguishing between stationary and nonstationary components proves difficult, Juselius (2006) suggests several formal and informal procedures to determine the rank. They include (a) the trace test, (b) the modulus of the roots of the companion matrix, (c) the graphical visualization of the recursively calculated trace test statistics, and (d) the graphical inspection of the stationarity of the cointegrating relations.

The trace test uses the likelihood ratio principle and tests the null hypothesis that at most \( r \) cointegration vectors exist against a general (unrestricted) alternative hypothesis that more than \( r \) cointegration vectors exist. The trace statistic is calculated as follows:

\[
Q_r = -T \sum_{i=r+1}^{p} \ln(1 - \hat{\lambda}_i),
\]

(11)

where \( T \) is the sample size, \( p \) is the dimension of the vector, and \( \hat{\lambda}_{r+1}, \ldots, \hat{\lambda}_p \) are the ordered eigenvectors obtained from the generalized eigenvalue problem as described in Johansen and Juselius (1990).

Although the trace statistic \( Q_r \) exhibits a nonstandard distribution, Johansen (1988,
1995) and Osterwald-Lenum (1992), among others report asymptotic critical values. Two problems exist with these critical values. First, the asymptotic distribution of $Q_r$ with a small sample size provides a generally poor approximation of the true distribution. Juselius (2006) shows that in such case the test experiences substantial size and power distortions. To address this problem, we use the Bartlett small sample correction (Bartlett, 1937) of the trace test due to Johansen (2002). Second, Nielsen (2004) shows that the number and location of shift dummy variables affect the asymptotic distribution of the test, and as a result, we cannot use the conventional critical values to determine the cointegration rank. To address this second problem, we simulate new asymptotic critical values using the simulation program in CATS in RATS version 2 with 1000 random walks and 10,000 replications.

Table 4 provides trace test evidence for rank determination. We report the estimated eigenvalues, the trace statistics $Q_r$, the Bartlett corrected trace statistic $Q_r^*$ and the simulated 95% quantile $C_{0.95}^D$, taking into account the presence of the shift dummy variable and the $p$-values for $Q_r$ and $Q_r^*$ ($p$-value and $p$-value*, respectively). Two points are noteworthy: (i) the Bartlett correction lowers the trace test statistic, and, (ii) $C_{0.95}^D$ is lower than $C_{0.95}^C$, which does not account for the shift dummy variable.  

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6 To check the robustness of this result, we compute the trace test statistic excluding the last part of the sample, which includes the period of the financial crisis. We construct two sub-samples. The first uses the Lehman Brothers bankruptcy of September 2008 as the ending date of the sub-sample, while the second uses the beginning of the sub-prime crisis of October 2007 as the ending date of the sub-sample. The finding of cointegration rank $r = 2$ continues to hold in both cases. We also computed the trace test statistic excluding the transition period of the euro (January 1999-December 2001). The finding of cointegration rank $r = 2$ holds, once again. Therefore, the implications of our analysis do not hinge on the presence of the crisis data in the sample nor on the data of the of the transition period. The details are available from the authors. We also compute the trace test using a different model specification that excludes all deterministic terms except the centered seasonal dummy variables. The test results overwhelmingly favor the choice of $r = 2$, with a $p$-value of about 0.8.
The first two estimated eigenvalues appear to differ from zero. A substantial gap also exists between the second and the third eigenvalue, which points to a cointegrating rank of two. The trace test $Q_r$ and the Bartlett small-sample adjusted trace test $Q'^r$ compared with the simulated $C^{1.0}_{0.95}$ confirm this conjecture. The tests reject the nulls of zero and one cointegrating rank. In contrast, the trace test and the Bartlett small-sample adjusted trace test cannot reject the null of $r = 2$ cointegrating relationships and, therefore, $p - r = 3$ common trends in the model. Rejecting the hypothesis that $r = 1$ runs counter to the frequent use of single equation models in the exchange rate determination literature. That is, a single equation model implies just one long-run cointegrating relationship between the relevant variables, whereas concluding that $r = 2$ means that existing data require a more complex model.

### 3.4 Further evidence for rank determination

Determining the cointegrating rank possibly proves as one of the most difficult tasks in empirical work using non-stationary data. Juselius (2006) strongly cautions against the exclusive dependence on trace-test evidence, as the LR tests may possess low power when the eigenvalue comes close to the non-stationary boundary. Hence, we should make use of as much additional information as possible for this purpose, as discussed by Juselius (2006). In this section, following Juselius (2006), we consider additional rank-relevant evidence, including the modulus of the characteristic roots of the model, the recursive graphs of the trace statistic, and the graphs of the cointegrating relations.

The calculus of the roots of the companion matrix complements the information of the

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7 This conclusion is only tentative because of the unknown sampling distribution of the eigenvalues, which precludes testing whether eigenvalues significantly differ from one.

8 The presence of multiple cointegration vectors indicates that an equilibrium subspace exists rather than a unique equilibrium relationship to which the system adjusts. That means that the variables tie together in different ways in the long-run.

9 The choice of the appropriate rank is critical because all the subsequent results depend on that choice.
rank test. Table 5 shows that for \( r = 2 \), the modulus of the largest unrestricted root drops to 0.562.\(^{10}\) No root over 1 suggests that the model does not include any explosive root and that all the eigenvalues, apart from the imposed unit roots, distinctly differ from unity. The restriction \( r = 2 \) seems an appropriate choice of reduced rank for the error-correction term. This finding matches the results of the trace test, but is only indicative because the roots do not come with confidence bands.

The recursive trace estimation uses a forward procedure based on Hansen and Johansen (1999), which visually displays the progression of the long-term linkages among the variables in the cointegrating system, and examines the sample dependence of the estimated cointegrating rank. This, in turn, enables the investigation of the robustness of the results for different sample sizes. We estimate the parameters based on a subsample from \( t = 1, \ldots, T_0 \). The recursive nature of the process involves adding one observation at a time to generate samples \( t = 1, \ldots, T_0 + t \), where \( t = 1, \ldots, T - T_0 \). Two different procedures exist, the X(t)- and R1(t)-forms.\(^{11}\) In the X(t)-form, we re-estimate all the parameters during the recursions, while in the R1(t)-form, we only re-estimate the long-term parameters, holding the short-term estimates fixed.

Figure 6 shows the time series graph of the trace statistic derived from the recursive process, where the upper panel show the results in the X(t)-form and the lower panel show the results in the R1(t)-form. Using an expanding window,\(^{12}\) we calculate the trace test statistic adding one observation at a time (Hansen and Johansen, 1999) and then divide the test statistics by its 5-percent critical value. Following Juselius (2006), if the cointegrating rank is \( r < p \), then

\(^{10}\) If we impose \( r = 3 \) when the appropriate rank is \( r = 2 \), then the third root comes closer to unity, and we should reduce \( r \) from 3 to 2. When \( r = 2 \), we observe the lowest first root beyond the unit root is 0.562. See Juselius (2006).

\(^{11}\) By fixing the estimates of the short-run parameters, we reduce the variance of the long-run parameters, which is the primary interest of cointegration analysis (Hansen and Johansen, 1999). This motivates the R1(t)-form.

\(^{12}\) We fix the base period, January 1999 to December 2002, at about 35 percent of the sample, following the suggestion of Brüggemann, Donati, and Warne (2003).
the recursively calculated trace statistics for \( j = 1, \ldots, r \) should display a value above one throughout the sample and grow linearly but shall stay constant for \( j = r + 1, \ldots, p \). As both panels of Figure 14 reveal, the recursively calculated trace statistics exhibit a linear growth for \( j = 1 \) and 2, but no growth for \( j = 3, 4, \) and 5. The first two linearly growing trace statistics correspond to two cointegration relations, which supports the choice of \( r = 2 \), while the remaining three relations indicate small eigenvalues, which correspond to a unit root or near-unit root.

Finally, we inspect the graphs of the five cointegrating relationships for evidence of stationarity. In our case, the first two relations appear stationary, and the remaining three do not. Such evidence supports the choice of rank \( r = 2 \) (Juselius, 2006). Figures 7 through 11 graph the individual cointegrating relationships of the unrestricted model. The upper panel of each graph shows the given cointegration relation based on the X(t)-form and the lower panel shows the same relationship based on the R1(t)-form. The order of cointegration is that of decreasing stationarity. We also notice that no trend exists in the first and second cointegrating relationships, as neither interest rates, \( ppp_t \), nor inflation rates should contain a deterministic trend, and their variance suggests stability over the sample period. The graphs of the third, fourth, and fifth relationships show a persistent behavior and do not suggest mean-reverting dynamics. On the other hand, the graphs of the first and second cointegrating relationships repeatedly cross the mean, which suggest stationarity, supporting the results of the trace tests.

The validity of our statistical inferences requires that no I(2) variables should enter the model. Juselius (2006) and Juselius and Toro (2005) suggest various criteria to investigate the presence of I(2) variables. First, they suggest comparing the trace test to the Bartlett corrected trace test. The ratio between \( Q_r \) and \( Q^*_r \) should fall between 1 and 1.2 and should not exceed
1.5. Second, the first unrestricted root of the companion matrix should not jump up close to unity, which indicates the presence of I(2) variables in the system. Third, graphs of the X(t)-form and the R1(t)-form look different if the data contain some I(2) variables. Applying these criteria to the data shows that no I(2) trends exist.

3.5 Tests of parameter constancy

A well-specified model exhibits parameter constancy. This proves particularly important during a financial crisis. Hansen and Johansen (1999) suggest applying a fluctuation test to the nonzero eigenvalues of the reduced-rank matrix. The test provides general information with regard to the constancy of the parameters because we can express the eigenvalues as quadratic functions of the \( \alpha \) and \( \beta \) parameters (Juselius, 2006). If both are constant, the eigenvalues will share this property. The fluctuation test rejects parameter constancy when the recursively calculated eigenvalues fluctuate excessively. We apply the test to the eigenvalues themselves, \( \lambda_i \), and to the transformation \( \xi = \log(\lambda_i/(1 - \lambda_i)) \) to obtain a symmetrical representation of their limiting distribution. We can also jointly evaluate the constancy of the eigenvalues by considering the sum of the transformed eigenvalues.

Figure 12 reports the time path of the two largest eigenvalues with their 95-percent confidence bands computed using the Bartlett kernel estimator of the asymptotic variance. These two eigenvalues significantly differ from zero for the entire sample, reinforcing the evidence in favor of two cointegrating vectors. Figure 13 displays the time path of the transformed eigenvalues and their sum, while Figure 14 presents the corresponding fluctuation tests. In the R1(t)-form, we do not reject the constancy of \( \xi_1, \xi_2 \), and their sum because the test statistic falls below the 5-percent critical value. Conversely, the test in the X(t)-form reject the constancy of \( \xi_1, \xi_2 \), and their sum due to the fluctuations in the beginning of the recursion. We can ascribe
this outcome, however, to the small sample size (Juselius, 2006) or, more importantly, to the euro transition period.\footnote{During the transition period, which lasts from January 1999 to December 2001, transactions in the countries of the euro area could use both the euro and national currencies. During this transition period, the euro only serves as an accounting unit, and euro notes and coins only start circulating in January 2002, when countries withdraw their national currencies from circulation.}

We further investigate whether any significant structural break exists in the cointegration vectors using two additional recursively calculated tests developed by Hansen and Johansen (1999): the test for the constancy of $\beta$ and the test for the constancy of the log-likelihood function.

The test for the constancy of $\beta$ investigates parameter constancy of the cointegration space. The null hypothesis of the test states that the cointegration vectors estimated over the full sample do not differ from the cointegration vectors estimated recursively. The test statistic is asymptotically distributed as chi square with $r \times r$ degrees of freedom. Under the null hypothesis of constancy of $\beta$, the 95-percent quantile of the test is 18.3.

The test for the constancy of the log-likelihood function investigates parameter constancy of the model. The test is similar to the recursive Chow test used in single equation models (Juselius, 2006). Under the null hypothesis of constant parameters, the 95-percent quantile of the test is 1.36. Figures 15 and 16 illustrate the test results. Figure 15 shows that stability of $\beta$ may not exist for the X(t)-form, but does exist for the R1(t) form. Instability in the X(t)-form, however, occurs mainly at the beginning of the recursion and for a brief period of time in the first quarter of 2005, which coincides with the start of the low interest rate policy of the European Central Bank and the weakening of the euro against the pound, and in the middle of 2008, which approximately corresponds to the beginning of the global financial crisis. Juselius (2006) recommends placing more reliance on the R1(t)-form plot, and, based on these
considerations, we find evidence that suggests relatively stable parameters of the cointegration space. Similar conclusions emerge by considering the time path of the log-likelihood function in Figure 16.

3.6 Tests of long-run exclusion and long-run weak exogeneity.

This subsection tests for long-run exclusion and weak exogeneity,\(^{14}\) which provides useful information on the relevance and the differential role of the variables in the cointegrating space.

Long-run exclusion tests investigate whether we can exclude any variables from the cointegration space. We formulate the tests as a zero row in the \( \beta \)-matrix (i.e., \( \beta_{ij} = 0 \), \( j = 1, \ldots, r \)). Table 6, Panel A, reports the test statistics, asymptotically distributed as \( \chi^2(r) \),\(^{15}\) which show that for \( r = 2 \), we cannot exclude any of the variables, including the shift dummy \( C_{2007:10} \) and the constant. This important result signifies that all the variables participate in the cointegration space and enter the long-run relationships. The significance of the restricted shift dummy variable is of particular interest, since it accounts for a change in the equilibrium means of the cointegrating relations in 2007:10, associated with the recent financial crisis.

Tests for weak exogeneity examine whether leading or driving forces exist in the systems in the long-run. A variable exhibits weak exogeneity when it significantly influences the remaining variables in the error-correction process, but is not significantly affected by those other variables in the long-run adjustment process. In other words, a weakly exogenous variable dominates and plays a leading role in the system. We formulate these tests as a zero row in the \( \alpha \)-matrix (i.e., \( \alpha_{ij} = 0 \), \( j = 1, \ldots, r \)). This means that a variable does not respond to any of the (long-run) error-correction terms and, thus, we consider it as weakly exogenous with respect to

\(^{14}\) Juselius (2006) discusses these tests in detail.

\(^{15}\) Whereas likelihood ratio testing for cointegrating rank leads to a nonstandard inference situation, conditional likelihood ratio testing, for a given cointegrating rank, produces standard asymptotically chi-squared test statistics.
the long-run parameters $\beta'$. If we do not reject the null hypothesis, then we can say that the variable in question “drives” (common stochastic trend) the system: it “pushes” the system, but it is not “pushed” by it. It also means that the sum of the cumulated empirical shocks to the variable in question defines one common driving trend. Table 6 reports the likelihood ratio test statistics, asymptotically distributed as $\chi^2(r)$. For $r = 2$, we reject the weak exogeneity hypothesis for the UK and the German inflation rates at any conventional level. We fail to reject, however, the hypothesis for the PPP condition ($p$-value = 0.211), the UK interest rate $i_t^*$ ($p$-value = 0.138), and the German interest rate $i_t$ ($p$-value = 0.062), although the latter case is borderline. Considering the German interest rate weakly exogenous, however, proves consistent with the choice of the rank $r = 2$. These findings are quite similar to the findings in Juselius and MacDonald (2000, 2004): inflation rates adjust to interest rates and the real exchange rate and not vice versa. As emphasized by Juselius and MacDonald (2004), this empirical result contradicts the predictions of models of exchange rate dynamics built on rational expectations and UIP, and in particular provides evidence against the Fisher hypothesis in open economies. We can justify this result, however, by appealing to the “imperfect knowledge economics” approach developed by Frydman and Goldberg (2003, 2006), which shows that under imperfect information expectations, exchange rates fluctuations do not represent movements toward a fundamental purchasing power equilibrium, but movements generated by traders’ behavior in the foreign exchange market (Juselius and MacDonald, 2004).

### 3.7 Linear restrictions on the cointegration space

This subsection explores the existence of valid restrictions on the cointegration space in the I(1)
cointegrated VAR(3) model under the restriction of two cointegrating vectors. The tests include the shift dummy variable $C_{2007:10}$, and the restricted constant. Table 7 enumerates the test results. First, we report test results for the stationarity of $\Delta \pi_t$ ($H_1$), the inflation differentials ($H_2$), the interest rate differentials ($H_3$), and the real interest rate in Germany and the UK ($H_4$ and $H_5$, respectively). We find evidence of stationarity only for the inflation differentials, (although the $p$-value is not very high) when the rest of the cointegrating vectors remain unrestricted. The stationarity of inflation differentials, however, is somewhat ambiguous, since the unit-root tests refute stationarity. Next, we report test results for the stationarity of the real interest rate differentials, imposing ($H_6$) and not imposing ($H_7$ and $H_8$) the full proportionality restrictions. We fail to find evidence of stationarity except, marginally, for $H_7$. We then test if inflation differentials ($H_9$), the interest rate differentials ($H_{10}$), the real interest rates in Germany and the UK ($H_{11}$ and $H_{12}$, respectively), and the Fisher parity conditions ($H_{13}$) combine with $\Delta \pi_t$ to produce linear stationary relationships. We find evidence of stationarity only when we combine the inflation rate differentials with $\Delta \pi_t$. We, therefore, identify the first cointegration vector where $\Delta \pi_t$ ($H_1$) and the inflation differentials ($H_2$) cointegrate. That is,

$$\Delta \pi_t - \Delta \pi_t^* + 0.502 \Delta \pi_t + 0.001 C_{2007:10} + 0.002 \sim I(0). \quad (12)$$

In the long-run, inflation differentials link to deviations from PPP. Equation (12) clearly implies that although $\Delta \pi_t$ is not by itself a stationary process, it becomes stationary when combined with $\Delta \pi_t$.

16 The hypotheses are of the form $H = (H\phi, \psi^*)$ where $H$ is the design matrix, $\phi$ contains the restricted parameters, and $\psi$ is a vector of parameters that are freely estimated. For details see Juselius (2006).

17 Since the rank equals two ($r = 2$), we can only test for cointegration, where theoretical relations restrict at least two of the parameters. This requirement is imposed by the degrees of freedom of the $\chi^2(\nu)$ distribution, which are calculated as $\nu = k - (r - 1)$, where $k$ is the number of restrictions (Juselius, 2006).
with the inflation rate differentials.\textsuperscript{18} We note that PPP in conjunction with unrestricted inflation rates (\( H_{14} \)) also yields a stationary outcome. This implies non-symmetric adjustment costs. The \( p \)-value associated with \( H_{14} \), however, is lower than the \( p \)-value associated with \( H_9 \) and a LR test (\( \chi^2(1) = 1.14 \), \( p \)-value = 0.285) confirms the validity of the symmetry restriction.\textsuperscript{19} The coefficient on \( ppp_t \) implies that a 1-percent change in \( ppp_t \) leads, in the long-run, to approximately 0.5-percent cost of adjustment in the inflation rate differentials. We do not find, however, evidence of stationarity when we combine the interest rate differentials with \( ppp_t \).

These results strikingly differ from those of Johansen and Juselius (1992), Juselius (1995), and Juselius and MacDonald (2000, 2004), who finds that \( ppp_t \) becomes stationary only when combined with the interest rate differential and conclude that capital markets and commodity markets are interdependent.

Finally, we find evidence of a stationary relation when we combine the inflation and interest rates in the Germany with the UK inflation rate (\( H_{15} \)). We therefore identify the second cointegration vector where the German interest rate cointegrates with the inflation rates in Germany and the UK. That is,

\[
(i_t - \Delta p_t) + 2.175 \Delta p_t - 0.003 C_{2007:10} - 0.005 \sim I(0) \tag{13}
\]

\textsuperscript{18} This result questions the stationarity of the inflation rate differential (\( H_2 \)). That is, if the inflation rate differential is really I(0), then it cannot cointegrate with \( ppp_t \), which is I(1).

\textsuperscript{19} For completeness, we also test, following Pedersen (2002b), for cointegration between \( ppp_t \) and the rate of inflation of Germany or the UK separately, which implies that the adjustment costs are borne unilaterally by only one country. In each case, we reject the hypotheses that \( ppp_t \) forms a stationary relation with the German inflation rate alone (\( \chi^2(2) = 11.739, p \)-value = 0.003) or with the UK inflation rate alone (\( \chi^2(2) = 5.264, p \)-value = 0.072) at the 5-percent level.
A test of the joint stationarity of $H_9$ and $H_{15}$ yields a $\chi^2(4)$ statistic of 7.393 with an associated $p$-value of 0.1119, which indicates that the two cointegrating relations span the entire cointegration space.

Tables 8, 9, and 10 report a structural representation of the cointegration space containing all the information included in the restrictions (i.e., the estimates of $\alpha$, $\beta$, and $\Pi$ matrices subject to the rank condition that $r = 2$, with $\beta_1'$ and $\beta_2'$ normalized for $\Delta p_i$ and $\Delta p_i^*$, respectively, the structural restrictions defined by $H_9$ and $H_{15}$, and weak exogeneity\textsuperscript{20}). Note that the joint estimation of the cointegrating vectors (i.e., the $\beta$ matrix) alters slightly the values of the unconstrained coefficients when compared to equations (12) and (13). The estimates of the $\alpha$ matrix suggest that the two cointegrating relations adjust significantly in the German and UK inflation rates. This conforms to the weak exogeneity of the German and UK inflation rates. The row of Table 10 gives the estimates of the combined effect of the two cointegrating relations. The German inflation rate exhibits significant responses to itself, the PPP condition, and the German interest rate. The UK inflation rate, on the other hand, exhibits significant responses to itself, the PPP condition, and the German interest rate. Juselius and MacDonald (2004) find what they call the “price puzzle effect” (i.e., inflation rates do not affect nominal interest rates, whereas nominal interest rates positively affect inflation rates). Our findings differ, in part, from Juselius and MacDonald (2004). We find that nominal interest rates exert a negative effect on inflation rates and the effect is limited to the German inflation rate. The UK interest rate does not affect inflation in Germany or the UK. Finally, we note that the shift dummy variable $C_{2007:10}$ also exerts a significant effect on inflation rates.

\textsuperscript{20} The unrestricted estimates of the $\alpha$ matrix, obtained without imposing the weak exogeneity restriction, are not significantly different from zero.
3.8  Analysis of the common stochastic trends

In this section, we estimate the moving average representation of the cointegrated system in order to extract information about the nonstationary components that drive the system in the long run. The moving-average representation of equation (1) is given by:

\[ x_t = C \sum_{i=0}^{L} (\varepsilon_i + \Psi D_i) + C_1(L)(\varepsilon_i + \phi D_i) + x_0, \]

(14)

where \( C = \beta_\perp (\alpha_\perp \Psi \beta_\perp)^{-1} \beta_\perp \alpha_\perp' \) equals the long-run impact matrix.

We report the common-trend representation corresponding to the restricted VAR model \((r = 2)\) subject to the weak exogeneity conditions on \( i_t, i_t^*, \) and \( ppp_t \) imposed on \( \alpha \) and the restrictions imposed on \( \beta \) by \( H_9 \) and \( H_{15} \). By construction, the three common stochastic trends in the system equal the cumulated shocks of \( i_t, i_t^* \), and \( ppp_t \) (i.e., \( \sum \varepsilon_{i_t}, \sum \varepsilon_{i_t^*}, \) and \( \sum \varepsilon_{ppp_t}, \) respectively). Tables 11, 12, and 13 report the estimates of \( \alpha_\perp \), the associated loadings (weights) \( \beta_\perp \), and the estimates of the long-run impact matrix. Table 11 identifies each the three common trends.

Table 12 highlights the dynamics of the system. We observe that the first common trend, identified by the cumulated shocks of the German interest rate, affects the rates of inflation in Germany and in the UK (as well as itself). On the other hand, the second common trend, identified by the cumulated shock of the UK interest rate, affects only itself. Finally, the third common trend, identified by the cumulated shocks of \( ppp_t \), affects both the inflation in Germany and the UK (as well as itself). Table 13 displays the long-run effects of a shock to the system. The significance of each element of the \( C \) matrix provides an indication of the effect of a shock on each of the variables in the system. We note that the cumulative shocks to \( \Delta p_t \) and \( \Delta p_t^* \) are,
by construction, equal to zero, since the UK and German inflation rates only adjust to the rest of the variables in the system. On the other hand, cumulative shocks to the German interest rate affect the UK and German inflation rates, in addition to itself. Similarly, shocks to the PPP condition significantly affect the UK and German inflation rates, in addition to itself. Shocks to the UK interest rate only affect itself.

4. Conclusion

This paper examines the empirical validity of the PPP hypothesis between the UK and the Euro Area, represented by Germany, the largest of its members. Following Juselius (1995) and Juselius and MacDonald (2000, 2004), we use the error-correction and moving-average representations of a five-dimensional VAR model, whose elements include the German and UK long term interest rates, the German and UK inflation rates, and the real exchange rate. The analysis uses monthly data from the introduction of the euro in January 1999 through April 2011.

The econometric analysis does not validate the linear restrictions implied by the Johansen and Juselius’ (1992) original hypothesis, subsequently confirmed by Juselius (1995) and Juselius and MacDonald (2000, 2004), among others. That is, these authors conclude that stationarity of $ppp_i$ occurs when we link PPP to UIP. We do not find that the nonstationarity of the PPP condition associates with the nonstationarity of the interest rates differentials to produce a stationary relation. This conclusion, although surprising, is not totally unexpected, given the historical resistance of the UK to engage in a deep financial integration and commitment with the European Union, and the diverging and asynchronous business cycles of the UK and the Euro Area.

On the other hand, we do not reject the PPP hypothesis. We find that two valid cointegrating relationships do exist. First, the PPP condition cointegrates with the inflation rates differentials. The Germany-UK PPP relation is not stationary by itself; but, stationarity emerges
when we link PPP to the inflation rates of both countries. This weak support for PPP, however, provides encouragement, given that (i) the monthly frequency of the data does not favor PPP (Hakkio and Rush, 1991), which is a long-run phenomenon, and (ii) the sample includes the transition period of the introduction of the euro, which a variety of exogenous elements likely contaminates, and the global financial crisis, which prompts aggressive interest rate policies in the months following the collapse of Lehman Brothers. Second, the UK and German inflation rates cointegrate with the German interest rate. Pedersen (2002) defines PPP with adjustment and uses adjustment costs to develop his specification, where the inflation rate differential represents the adjustment cost. In any event, this empirical result contradicts the predictions of conventional monetary models of exchange rate dynamics built on rational expectations and interest rate parity.

Juselius and MacDonald (2000, 2004) find that long-term interest rates and the PPP condition encompass the weakly exogenous variables. Further, they find that the inflation rates do not drive the rest of the system, but rather respond to the variables in the rest of the system. We find similar results.

First, we find that the German and UK inflation rates exhibit equilibrium-adjusting behavior (i.e., they are “pulled” back to equilibrium when they are “pushed” away from it). The German and UK interest rates, as well as the PPP condition, on the other hand, are weakly exogenous to the system (i.e., they affect the stochastic behavior of the German and UK inflation rates without being affected by them).

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21 This argument is also made by Manzur and Chan (2010).

22 We can also justify the empirical support of PPP induced by the nonstationarity of inflation differentials by appealing to models where foreign exchange traders’ “imperfect information expectations” (Frydman and Goldberg, 2003, 2005) generate movements in exchange rates or models of firms in imperfectly competitive markets facing inflation costs (Bacchiocchi and Fanelli, 2005). See also Frydman, Goldberg, Johansen, and Juselius (2012) in relation to “imperfect information expectations.”
Second, we find that the system is “pushed” by three common trends, associated with the cumulated shocks to the German and UK interest rates and the PPP condition, with the first common trend, associated with the cumulated shocks of the German interest rate and third common trend, associated with the cumulated shock of the PPP condition playing a dominant role, as each drives themselves as well as the inflation rates in Germany and the UK, while the second common trend, associated with cumulated shocks of the UK interest rate, drives only itself. In this regard, we can say that Germany dominates the UK in the cointegrating relationship in that the UK interest rate does not drive any other variable, whereas the German interest rate drives both the German and UK inflation rates, perhaps reflecting the “safe haven” role of the German financial markets in the European Union. A similar interpretation is suggested by Juselius and MacDonald (2004) in the context of the role of the dollar as world reserve currency.

References:


Table 1: Augmented Dickey-Fuller unit root tests

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<td>$(i_t - \Delta p_t) - (i_t^* - \Delta p_t^*)$</td>
<td>-1.884</td>
<td>12</td>
<td>-8.581</td>
<td>11</td>
</tr>
</tbody>
</table>

Note: The test critical values at the 0.01 and 0.05 levels are –3.477 and –2.881, respectively. $k$ is the number of lags. The Akaike information criterion ($AIC$) selects the number of lags.

Table 2: Dickey-Fuller-GLS unit root tests

<table>
<thead>
<tr>
<th>Variable</th>
<th>Levels</th>
<th>k</th>
<th>Differences</th>
<th>k</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\Delta p_t$</td>
<td>-1.843</td>
<td>11</td>
<td>-3.735</td>
<td>12</td>
</tr>
<tr>
<td>$\Delta p_t^*$</td>
<td>-1.453</td>
<td>12</td>
<td>-2.698</td>
<td>3</td>
</tr>
<tr>
<td>$i_t$</td>
<td>-1.649</td>
<td>1</td>
<td>-4.480</td>
<td>1</td>
</tr>
<tr>
<td>$i_t^*$</td>
<td>-1.643</td>
<td>7</td>
<td>-4.107</td>
<td>2</td>
</tr>
<tr>
<td>$ppp_t$</td>
<td>-0.812</td>
<td>1</td>
<td>-3.416</td>
<td>3</td>
</tr>
<tr>
<td>$\Delta p_t - \Delta p_t^*$</td>
<td>-0.708</td>
<td>12</td>
<td>-2.924</td>
<td>4</td>
</tr>
<tr>
<td>$i_t - i_t^*$</td>
<td>-1.683</td>
<td>0</td>
<td>-4.517</td>
<td>2</td>
</tr>
<tr>
<td>$i_t - \Delta p_t$</td>
<td>-1.480</td>
<td>11</td>
<td>-8.451</td>
<td>4</td>
</tr>
<tr>
<td>$i_t^* - \Delta p_t^*$</td>
<td>-0.842</td>
<td>12</td>
<td>-2.872</td>
<td>4</td>
</tr>
<tr>
<td>$(i_t - \Delta p_t) - (i_t^* - \Delta p_t^*)$</td>
<td>-0.932</td>
<td>12</td>
<td>-4.706</td>
<td>2</td>
</tr>
</tbody>
</table>

Note: The test critical values at the 0.01 and 0.05 levels are –2.582 and –1.943, respectively. $k$ is the number of lags. The Akaike information criterion ($AIC$) selects the number of lags.
Table 3: Specification tests (full rank model)

<table>
<thead>
<tr>
<th>Tests for autocorrelation</th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>LM(1): ( \chi^2 ) (25)</td>
<td>38.357</td>
<td>[0.043]</td>
<td></td>
</tr>
<tr>
<td>LM(2): ( \chi^2 ) (25)</td>
<td>36.967</td>
<td>[0.058]</td>
<td></td>
</tr>
</tbody>
</table>

**Test for Normality**

<p>| | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>( \chi^2 ) (10)</td>
<td>12.020</td>
<td>[0.284]</td>
</tr>
</tbody>
</table>

**Test for ARCH**

<p>| | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>LM(1): ( \chi^2 ) (225)</td>
<td>282.807</td>
<td>[0.005]</td>
<td></td>
</tr>
<tr>
<td>LM(2): ( \chi^2 ) (450)</td>
<td>512.408</td>
<td>[0.022]</td>
<td></td>
</tr>
</tbody>
</table>

**B. Univariate tests**

<table>
<thead>
<tr>
<th>ARCH(3)</th>
<th>Normality</th>
<th>( R^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \Delta^2 p_t )</td>
<td>4.192</td>
<td>[0.241]</td>
</tr>
<tr>
<td>( \Delta^2 p_t^* )</td>
<td>0.307</td>
<td>[0.959]</td>
</tr>
<tr>
<td>( \Delta p_{pp} )</td>
<td>0.648</td>
<td>[0.885]</td>
</tr>
<tr>
<td>( \Delta t_i )</td>
<td>1.535</td>
<td>[0.674]</td>
</tr>
<tr>
<td>( \Delta t_i^* )</td>
<td>6.118</td>
<td>[0.106]</td>
</tr>
</tbody>
</table>

Note: The multivariate LM test statistic is asymptotically distributed as \( \chi^2 \) with \( p^2 \) degrees of freedom (Johansen, 1995). The multivariate and univariate normality tests are asymptotically \( \chi^2 \) distributed, with \( 2p \) degrees of freedom in the multivariate and \( 2p \) degrees of freedom in the univariate case, respectively (Doornik and Hansen, 2008). The multivariate ARCH test statistic is approximately distributed as \( \chi^2 \) with \( \frac{q}{4}p^2(p+1) \) degrees of freedom. The univariate ARCH test is approximately distributed as \( \chi^2 \) with \( k \) degrees of freedom (Dennis et al., 2006).

Table 4: Tests for cointegration rank

<table>
<thead>
<tr>
<th>(p - r)</th>
<th>r</th>
<th>( \lambda_i )</th>
<th>( Q_r )</th>
<th>( Q_r^{*} )</th>
<th>( C_{0.95}%^{D} )</th>
<th>p-value</th>
<th>p-value*</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>0</td>
<td>0.364</td>
<td>159.978</td>
<td>148.877</td>
<td>87.333</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td>4</td>
<td>1</td>
<td>0.313</td>
<td>94.898</td>
<td>89.150</td>
<td>63.677</td>
<td>0.000</td>
<td>0.001</td>
</tr>
<tr>
<td>3</td>
<td>2</td>
<td>0.158</td>
<td>40.789</td>
<td>38.568</td>
<td>42.883</td>
<td>0.078</td>
<td>0.124</td>
</tr>
<tr>
<td>2</td>
<td>3</td>
<td>0.078</td>
<td>16.076</td>
<td>15.232</td>
<td>25.772</td>
<td>0.476</td>
<td>0.543</td>
</tr>
<tr>
<td>1</td>
<td>4</td>
<td>0.030</td>
<td>4.321</td>
<td>4.126</td>
<td>12.023</td>
<td>0.669</td>
<td>0.696</td>
</tr>
</tbody>
</table>

Note: In addition to the number of common trends \( (p-r) \) and cointegrating vectors \( (r) \), this table reports the estimated eigenvalues \( \lambda_i \), the trace statistics \( Q_r \), the Bartlett small-sample corrected trace statistics, \( Q_r^{*} \), the 95-percent quantile from the asymptotic distribution corrected for deterministic component, \( C_{0.95}\%^{D} \), obtained using 2500 simulations, as well as the p-value of the test statistic and the p-value* of the Bartlett small-sample correction.
### Table 5: Modulus of the ten largest roots of the companion matrix of the VAR model

<table>
<thead>
<tr>
<th>r</th>
<th>r = 0</th>
<th>r = 1</th>
<th>r = 2</th>
<th>r = 3</th>
<th>r = 4</th>
<th>r = 5</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>0.938</td>
</tr>
<tr>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>0.874</td>
<td>0.904</td>
</tr>
<tr>
<td>1</td>
<td>1</td>
<td>0.562</td>
<td>0.569</td>
<td>0.659</td>
<td>0.654</td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>0.572</td>
<td>0.562</td>
<td>0.569</td>
<td>0.566</td>
<td>0.568</td>
<td></td>
</tr>
<tr>
<td>0.640</td>
<td>0.572</td>
<td>0.558</td>
<td>0.543</td>
<td>0.566</td>
<td>0.568</td>
<td></td>
</tr>
<tr>
<td>0.640</td>
<td>0.541</td>
<td>0.558</td>
<td>0.543</td>
<td>0.545</td>
<td>0.545</td>
<td></td>
</tr>
<tr>
<td>0.571</td>
<td>0.541</td>
<td>0.514</td>
<td>0.535</td>
<td>0.454</td>
<td>0.545</td>
<td></td>
</tr>
<tr>
<td>0.571</td>
<td>0.530</td>
<td>0.514</td>
<td>0.535</td>
<td>0.454</td>
<td>0.535</td>
<td></td>
</tr>
<tr>
<td>0.481</td>
<td>0.530</td>
<td>0.497</td>
<td>0.496</td>
<td>0.409</td>
<td>0.535</td>
<td></td>
</tr>
</tbody>
</table>

Note: The table shows the modulus of the estimated \( p \times k \) roots of the companion matrix from the VAR system, \( p \) is the number of variables and \( k \) is the number of lags of the VAR. The eigenvalues, \( \hat{\lambda}_i \) and the roots, \( \hat{\rho}_i \), relate to each other as follows: \( \hat{\lambda}_i = 1 - \hat{\rho}_i \). For example, \( \hat{\lambda}_i = 0 \) implies \( \hat{\rho}_i = 1 \), which corresponds to a unit-root process.

### Table 6: Tests of Long-Run Exclusion and Weak Exogeneity

<table>
<thead>
<tr>
<th>Panel A: Tests of Long-Run Exclusion</th>
<th>( r )</th>
<th>df</th>
<th>5% C.V.</th>
<th>( \Delta \rho_i )</th>
<th>( \Delta \rho_i^* )</th>
<th>( i_i )</th>
<th>( i_i^* )</th>
<th>( \text{ppp}_i )</th>
<th>( C_{2007:10} )</th>
<th>\text{constant}</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>2</td>
<td>2</td>
<td>5.991</td>
<td>40.036</td>
<td>32.959</td>
<td>11.738</td>
<td>6.985</td>
<td>8.726</td>
<td>10.737</td>
<td>7.538</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>[0.000]</td>
<td>[0.000]</td>
<td>[0.003]</td>
<td>[0.030]</td>
<td>[0.013]</td>
<td>[0.005]</td>
<td>[0.023]</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Panel B: Tests of Long-Run Weak Exogeneity</th>
<th>( r )</th>
<th>df</th>
<th>5% C.V.</th>
<th>( \Delta \rho_i )</th>
<th>( \Delta \rho_i^* )</th>
<th>( i_i )</th>
<th>( i_i^* )</th>
<th>( \text{ppp}_i )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>2</td>
<td>2</td>
<td>5.991</td>
<td>33.627</td>
<td>28.594</td>
<td>5.571</td>
<td>3.961</td>
<td>3.115</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>[0.000]</td>
<td>[0.000]</td>
<td>[0.062]</td>
<td>[0.138]</td>
<td>[0.211]</td>
<td></td>
</tr>
</tbody>
</table>

Notes: Numbers in brackets report \( p \)-values. The tests of long-run weak exogeneity restriction tests for the null hypothesis \( \alpha_i = 0 \). The tests are LR tests distributed as \( \chi^2(r) \).
### Table 7: Tests of cointegrating relations

<table>
<thead>
<tr>
<th></th>
<th>$\Delta p_t$</th>
<th>$\Delta p_t^*$</th>
<th>$i_t$</th>
<th>$i_t^*$</th>
<th>$ppp_t$</th>
<th>$C_{2007:10}$</th>
<th>constant</th>
<th>$\chi^2 (\nu)$</th>
<th>$p$-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$H_1$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>-0.002</td>
<td>0.004</td>
<td>36.091(3)</td>
<td>0.000</td>
</tr>
<tr>
<td>$H_2$</td>
<td>1</td>
<td>-1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0.002</td>
<td>0.000</td>
<td>5.512(3)</td>
<td>0.138</td>
</tr>
<tr>
<td>$H_3$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>-1</td>
<td>0</td>
<td>0.000</td>
<td>0.000</td>
<td>49.697(3)</td>
<td>0.000</td>
</tr>
<tr>
<td>$H_4$</td>
<td>1</td>
<td>0</td>
<td>-1</td>
<td>0</td>
<td>0</td>
<td>-0.001</td>
<td>0.002</td>
<td>22.230(3)</td>
<td>0.000</td>
</tr>
<tr>
<td>$H_5$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>-1</td>
<td>0</td>
<td>-0.003</td>
<td>0.003</td>
<td>12.820(3)</td>
<td>0.005</td>
</tr>
<tr>
<td>$H_6$</td>
<td>1</td>
<td>-1</td>
<td>-1</td>
<td>1</td>
<td>0</td>
<td>0.002</td>
<td>-0.001</td>
<td>9.050(3)</td>
<td>0.029</td>
</tr>
<tr>
<td>$H_7$</td>
<td>1</td>
<td>-1</td>
<td>0.06</td>
<td>-0.06</td>
<td>0</td>
<td>0.002</td>
<td>0.000</td>
<td>5.500(2)</td>
<td>0.064</td>
</tr>
<tr>
<td>$H_8$</td>
<td>1</td>
<td>-1.55</td>
<td>-1</td>
<td>1.55</td>
<td>0</td>
<td>0.003</td>
<td>-0.002</td>
<td>6.86(2)</td>
<td>0.032</td>
</tr>
<tr>
<td>$H_9$</td>
<td>1</td>
<td>-1</td>
<td>0</td>
<td>0</td>
<td>0.502</td>
<td>0.001</td>
<td>0.002</td>
<td>3.912(2)</td>
<td>0.203</td>
</tr>
<tr>
<td>$H_{9,4}$</td>
<td>-0.64</td>
<td>1.00</td>
<td>0</td>
<td>0</td>
<td>-0.352</td>
<td>-0.001</td>
<td>-0.002</td>
<td>2.051(1)</td>
<td>0.152</td>
</tr>
<tr>
<td>$H_{10}$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>-1</td>
<td>1.011</td>
<td>-0.002</td>
<td>0.005</td>
<td>30.803(2)</td>
<td>0.000</td>
</tr>
<tr>
<td>$H_{11}$</td>
<td>1</td>
<td>0</td>
<td>-1</td>
<td>0</td>
<td>-0.767</td>
<td>0.001</td>
<td>0.001</td>
<td>19.947(2)</td>
<td>0.000</td>
</tr>
<tr>
<td>$H_{12}$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>-1</td>
<td>-0.7</td>
<td>-0.002</td>
<td>0.002</td>
<td>10.327(2)</td>
<td>0.006</td>
</tr>
<tr>
<td>$H_{13}$</td>
<td>1</td>
<td>-1</td>
<td>-1</td>
<td>1</td>
<td>0.086</td>
<td>0.002</td>
<td>0.000</td>
<td>8.979(2)</td>
<td>0.011</td>
</tr>
<tr>
<td>$H_{14}$</td>
<td>-0.64</td>
<td>1.00</td>
<td>0</td>
<td>0</td>
<td>-0.352</td>
<td>-0.001</td>
<td>-0.002</td>
<td>2.051(1)</td>
<td>0.152</td>
</tr>
<tr>
<td>$H_{15}$</td>
<td>-1</td>
<td>2.175</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>-0.003</td>
<td>-0.005</td>
<td>0.775(2)</td>
<td>0.686</td>
</tr>
</tbody>
</table>

Notes: $\nu$ is the number of degrees of freedom, defined as $\nu = k - (r - 1)$, where $k$ is the number of restrictions.

### Table 8: A structural representation of the cointegration space

<table>
<thead>
<tr>
<th></th>
<th>$\Delta p_t$</th>
<th>$\Delta p_t^*$</th>
<th>$i_t$</th>
<th>$i_t^*$</th>
<th>$ppp_t$</th>
<th>$C_{2007:10}$</th>
<th>constant</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\hat{\beta}_1'$</td>
<td>1.000</td>
<td>-1.000</td>
<td>0.000</td>
<td>0.000</td>
<td>0.515</td>
<td>0.001</td>
<td>0.002</td>
</tr>
<tr>
<td>(NA)</td>
<td>(NA)</td>
<td>(NA)</td>
<td>(NA)</td>
<td>(3.715)</td>
<td>(1.899)</td>
<td>(3.340)</td>
<td></td>
</tr>
<tr>
<td>$\hat{\beta}_2'$</td>
<td>-0.468</td>
<td>1.000</td>
<td>0.468</td>
<td>0.000</td>
<td>0.000</td>
<td>-0.002</td>
<td>0.002</td>
</tr>
<tr>
<td>(-8.838)</td>
<td>(8.838)</td>
<td>(NA)</td>
<td>(NA)</td>
<td>(-5.436)</td>
<td>(-11.572)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: Numbers in parenthesis report $t$-statistics, where applicable.
### Table 9: Parameter estimates of the $\alpha$ matrix

<table>
<thead>
<tr>
<th></th>
<th>$\hat{\alpha}_1$</th>
<th>$\hat{\alpha}_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\Delta^2 p_t$</td>
<td>-2.423</td>
<td>-2.207</td>
</tr>
<tr>
<td></td>
<td>(-8.038)</td>
<td>(-6.530)</td>
</tr>
<tr>
<td>$\Delta^2 p_t^*$</td>
<td>-0.732</td>
<td>-1.743</td>
</tr>
<tr>
<td></td>
<td>(-2.608)</td>
<td>(-5.540)</td>
</tr>
<tr>
<td>$\Delta i_t$</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td></td>
<td>(0.000)</td>
<td>(0.000)</td>
</tr>
<tr>
<td>$\Delta i_t^*$</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td></td>
<td>(0.000)</td>
<td>(0.000)</td>
</tr>
<tr>
<td>$\Delta ppp_t$</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td></td>
<td>(0.000)</td>
<td>(0.000)</td>
</tr>
</tbody>
</table>

Notes: Numbers in parenthesis report $t$-statistics.

### Table 10: Parameter estimates of the $\Pi$ matrix

<table>
<thead>
<tr>
<th></th>
<th>$\Delta p_t$</th>
<th>$\Delta p_t^*$</th>
<th>$i_t$</th>
<th>$i_t^*$</th>
<th>$ppp_t$</th>
<th>$C_{2007:10}$</th>
<th>constant</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\Delta^2 p_t$</td>
<td>-1.391</td>
<td>0.216</td>
<td>-1.032</td>
<td>0.000</td>
<td>-1.248</td>
<td>0.002</td>
<td>0.000</td>
</tr>
<tr>
<td></td>
<td>(-7.616)</td>
<td>(1.276)</td>
<td>(-6.530)</td>
<td>(NA)</td>
<td>(-8.038)</td>
<td>(4.460)</td>
<td>(0.143)</td>
</tr>
<tr>
<td>$\Delta^2 p_t^*$</td>
<td>0.083</td>
<td>-1.011</td>
<td>-0.815</td>
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Notes: Numbers in parenthesis report $t$-statistics.
### Table 11: Estimates of the coefficients of the common trends: $\alpha_\perp$

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<td>(NA)</td>
<td>(NA)</td>
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Notes: Numbers in parenthesis report $t$-statistics.

### Table 12: Estimates of the loadings to the common trends: $\beta_\perp$

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<td>(-3.826)</td>
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<td>1.562</td>
<td>-0.265</td>
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<td>0.813</td>
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<td>(2.979)</td>
<td>(0.580)</td>
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<tr>
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<td>(0.873)</td>
<td>(7.585)</td>
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Notes: Numbers in parenthesis report $t$-statistics.
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<th>$\sum \varepsilon_{\Delta p_t^*}$</th>
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<th>$\sum \varepsilon_{i_t^*}$</th>
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Notes: Numbers in parenthesis report $t$-statistics.
Figure 1: Graphical analysis of the $\Delta p^t$ residuals

Figure 2: Graphical analysis of the $\Delta p^*_t$ residuals

Figure 3: Graphical analysis of the $ppp^t$ residuals
Figure 4: Graphical analysis of the $i_t$ residuals

Figure 5: Graphical analysis of the $i_t^*$ residuals
Figure 6: The recursively calculated trace test statistics

Figure 7: The first cointegrating relationship, $\beta_1'X_{1t}$ and $\beta_1'R_{1t}$
Figure 8: The second cointegrating relationship, $\hat{\beta}_2' \mathbf{X}_2t$ and $\hat{\beta}_2' \mathbf{R}_2t$

Figure 9: The third cointegrating relationship, $\hat{\beta}_3' \mathbf{X}_3t$ and $\hat{\beta}_3' \mathbf{R}_3t$
Figure 10: The fourth cointegrating relationship, $\hat{\beta}_4 X_{4t}$ and $\hat{\beta}_4 R_{4t}$

Figure 11: The fifth cointegrating relationship, $\hat{\beta}_5 X_{5t}$ and $\hat{\beta}_5 R_{5t}$
Figure 12: The time paths of the eigenvalues $\lambda_1$ and $\lambda_2$ with 95% confidence bands (dotted lines).

Figure 13: The time paths of the transformed eigenvalues $\xi_1$ and $\xi_2$ and their sum with 95% confidence bands (dotted lines).
Figure 14: Fluctuation tests of the transformed eigenvalues and their sum for the X(t)-form (dotted line) and the R1(t)-form (dashed line). The graphs are scaled by the 5% critical value (1.36) marked by the horizontal line.

Figure 15: Time path for the tests for $\beta_i$ equal to known $\beta$ scaled by the 5% critical value. The X(t)-form is represented by the dotted line and the R1(t)-form by the dashed line.
Figure 16: Time path of the log-likelihood function plotted for the X(t)-form (dotted line) and the R1(t)-form (dashed line).