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A panel cointegration analysis of the exchange rate pass-through

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Abstract

This paper investigates the presence of a long-run equilibrium relationship in the exchange rate pass-through (ERPT) equation for a panel of 27 OECD countries. Previous empirical panel data studies, such as Barhoumi [2005, "Exchange rate pass-through into import prices in developing countries: An empirical investigation", *Economics Bulletin*, Vol. 3 (26), 1–14.], have neglected the possibility of cross-sectional correlation and spillovers amongst countries. Since the strong economic and financial linkages between OECD countries cannot be ignored, we apply second generation panel unit root and panel cointegration tests which account for possible cross-section dependence across the units in the panel. Our results suggest the existence of a cointegrated equilibrium relationship between the variables in levels, as implied by the theoretical underpinning of the ERPT mechanism. When estimating the long-run pass-through coefficient, both FM-OLS and DOLS estimators show an incomplete pass-through, i.e. import prices sensitivity to exchange rate movements does not exceed 0.70% for our sample of OECD countries. This evidence of partial pass-through would represent a key element in understanding the ongoing global external imbalances.

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1 Introduction

The concept of the exchange rate pass-through (ERPT) refers to the degree of sensitivity of import prices to a one percent change in exchange rates in the importing nation's currency. Thorough knowledge of the degree of pass-through is of particular importance for several policy issues, including the design and the conduct of monetary policy, adjustment in trade balances, the international transmission of shocks and the optimal choice of exchange rate regime. Policymakers must be able to gauge how large the effects of currency movements are likely to be, in order to determine the appropriate monetary policy responses. Hence, it is an important task for empirical research to discriminate between the alternative models and econometric procedures in order to provide valid and reliable ERPT coefficients.

An important issue is the time-series properties of the data, particularly the non-stationarity and cointegration relationship, which must be considered with more accuracy. Failed to find an evidence of cointegration relationship in the data, several studies has estimated ERPT models in first differences where the information contained in "levels" variables is lost. Nevertheless, as predicted by the theoretical underpinning of the ERPT mechanism, a long-run or steady-state relationship between the key variables in levels, i.e. exchange rate and price series, should exist. Thus, using appropriate estimation techniques would help restore a cointegrated equilibrium relationship. Recently, there has been an increasing use of unit root and cointegration analysis in the context of panel data. As discussed in panel cointegration literature (see e.g. PEDRONI, 1999, 2001, 2004; BREITUNG and PESARAN, 2005, among others), conventional nonstationary tests have low power in small sample sizes, so adding the cross-section dimension to the time series dimension would increase the power of these tests.

Thus, due to recent developments in time series and panel data econometrics, some studies of the ERPT explicitly recognize the fact that exchange rate and price series are often non-stationary and may be cointegrated.¹ For example, in a panel of 24 developing countries, BARHOUMI (2005, 2006) gives support to the presence of a long-run equilibrium relationship between variables entering the ERPT equation. The author employed the popular first generation tests for non-stationary panel data: the IM et al. (2003, henceforth IPS) panel unit root tests as well as the residual-based tests for panel cointegration developed by PEDRONI (1999, 2004). Using the same approach, HOLMES (2006) were able to find a strong evidence of cointegrating relationship for 12 European Union countries. First generation tests, despite their advantages, are not exempt from critics. The key drawback of these tests is that the possible cross-sectional correlation and spillovers amongst countries are neglected. Indeed, the assumption of independence is usually is unrealistic, in particular in the analysis of macroeconomic or financial data which reveal strong inter-economy linkages.² Then, it is very restrictive to assume that some countries, such as European Union or OECD members, are independent and no economic or financial linkage does exist between them. From econometric point of view, PESARAN (2007) argued that not accounting for possible cross-sections dependence across units would entail a considerable size distortions in panels and may lead to flawed inference on the existence of the long-run relationship.

The goal of our paper is to overcome this mis-performance of first generation panel tests in the context of ERPT. We employ a class of panel unit root and panel cointegration tests which allow for serial correlation between the cross-sections, i.e. the so-called second generation tests. To the best of our knowledge, only DE BANDT, BANERJEE, and KOZLUK (2008) have applied these recent panel data techniques when investigating the issue of pass-through. In a sectoral study for 11 euro area countries, these authors show that the long-run relationship can be restored once the cross-section dependence is considered in the panel. In our study, unlike DE BANDT, BANERJEE, and KOZLUK (2008), we consider larger and more heterogeneous sample of countries, namely 27 OECD countries. The existence of

¹Although the number of those empirical works is still sparse.

²The occurrence of critical energy or economic events is in general common for several countries, such as oil price shocks.

strong inter-economic linkages between OECD countries cannot plausibly be ignored, consequently, we apply recently developed non-stationary panel techniques to tackle this issue. Also, we deal with the ERPT from a macroeconomic perspective using aggregate price measures, an issue of key importance for the conduct of monetary policy. Furthermore, second generation tests used in the context of our paper are different from DE BANDT, BANERJEE, and KOZLUK (2008). We employ the cross-sectionally augmented IPS panel unit root tests by PESARAN (2007) and the error-correction-based tests for panel cointegration by WESTERLUND (2007), which both account for possible cross-sectional dependencies. To our knowledge, no other study has applied these non-stationary panel methods in the context of ERPT. Finally, in order to provide valid and reliable long-run ERPT coefficients, we apply both FMOLS and DOLS group mean estimators, as introduced by PEDRONI (2001), to estimate the long-run ERPT. This approach is quite relevant since it allows the long-run cointegration relationships to be heterogeneous across countries.

The remainder of this paper is organized as follows. Section 2 describes the analytical framework that underlies our empirical specification and presents the data used in the study. In Section 3, we discuss the empirical methodology employed to test for stationarity and cointegration in panel data. Both FM-OLS and DOLS estimators of the long-run ERPT are presented in Section 4. Section 5 concludes.

2 Analytical framework and Data description

2.1 Pass-Through Equation

Our approach is to use the standard specification used in the pass-through literature as a starting point (see e.g. GOLDBERG and KNETTER, 1997; CAMPA and GOLDBERG, 2005). By definition, the import prices, MP_{it} , for any country i are a transformation of the export prices, XP_{it} , of that country's trading partners, using the nominal exchange rate, E_{it} (domestic currency per unit foreign currency):

$$MP_{it} = E_{it} \cdot XP_{it} \quad (1)$$

Using lowercase letters to reflect logarithms, we rewrite equation (1)

$$mp_{it} = e_{it} + xp_{it} \quad (2)$$

where the export price consists of the exporters marginal cost, MC_{it} and a markup, $MKUP_{it}$

$$XP_{it} = MC_{it} \cdot MKUP_{it} \quad (3)$$

In logarithms we have

$$xp_{it} = mc_{it} + mkup_{it} \quad (4)$$

So we can rewrite equation (2) as

$$mp_{it} = e_{it} + mc_{it} + mkup_{it} \quad (5)$$

Markup is assumed to have two components: (i) a specific industry component and (ii) a reaction to exchange rate movements

$$mkup_{it} = \alpha_i + \Phi e_{it} \quad (6)$$

Exporter marginal costs are a function of the destination market demand conditions, y_{it} , and wages in exporting country, w_{it}^* :

$$mc_{it} = \eta_0 y_{it} + \eta_1 w_{it}^* \quad (7)$$

Substituting (6) and (7) into (5), we derive

$$mp_{it} = \alpha_i + \underbrace{(1 + \Phi)}_{\beta} e_{it} + \eta_0 y_{it} + \eta_1 w_{it}^*, \quad (8)$$

The structure assumes unity translation of exchange rate movements. This empirical setup permits the exchange rate pass-through, represented by $\beta = (1 + \Phi)$, to depend on the structure of competition in one industry. Exporters of a given product can decide to absorb some of the exchange rate variations instead of passing them through to the price in the importing country currency. So if $\Phi = 0$, the pass-through is complete and their markups will not respond to fluctuations of the exchange rates. This is the case when import prices are determined in the exporter's currency (producer-currency pricing or PCP). And if $\Phi = -1$, exporters decide not to vary the prices in the destination country currency, thus they fully absorb the fluctuations in exchange rates in their own markups (LCP is prevailing).

Thus the final equation can be re-written as follows

$$mp_{it} = \alpha_i + \beta e_{it} + \gamma y_{it} + \delta w_{it}^* + \varepsilon_{it}, \quad (9)$$

The most prevalent result is an intermediate case where ERPT is incomplete (but different from zero), resulting from a combination of LCP and PCP in the economy. So, there is a fraction of import prices are set in domestic currency, while the remaining prices are set in foreign currency. Thus, the extent to which exchange rate movements are passed-through to prices will depend on the predominance of LCP or PCP: the higher the LCP, the lower the ERPT, and the higher PCP, the higher ERPT.

2.2 Data description

In this study, we consider the following panel of 27 OECD countries: Australia, Austria, Belgium, Canada, Czech Republic, Denmark, Finland, France, Germany, Greece, Iceland, Ireland, Italy, Japan, Korea, Luxembourg, Netherlands, Norway, New Zealand, Poland, Portugal, Slovak Republic, Spain, Sweden, Switzerland, United Kingdom and United States. The data are quarterly and span the period 1994:1-2010:4. Concerning our dependent variable, i.e. the domestic import prices, mp_{it} , we use the price of non-commodity imports of goods and services from OECD's Main Economic Outlook database. This represents import prices of core goods by excluding primary raw commodities because of their marked volatility.³ From the same database we take the real GDP as proxy for the domestic demand, y_{it} . To capture changes in foreign costs, we construct a typical export partners cost proxy, W_t^* , that used throughout the ERPT literature (see e.g. BAILLIU and FUJII, 2004; CAMPA and GOLDBERG, 2005): $W_t^* = Q_t \times W_t / E_t$, where Q_t is the unit labor cost based-real effective exchange rate, W_t is the domestic unit labor cost and E_t is the nominal effective exchange rate.⁴ Taking the logarithm we obtain the following expression: $w_{it}^* = q_t + w_t - e_t$. Since the nominal and real effective exchange rate series are trade weighted, we obtain a measure of foreign firms' costs with each partner weighted by its importance in the domestic country's trade. Data used to construct foreign producers costs - nominal effective exchange rate, consumer prices index and real effective exchange rate - are obtained from IMF's International Financial Statistics.

³As explained by IHRIG, MARAZZI, and ROTHENBERG (2006), when it was not possible to find import prices of core goods that exclude all primary raw commodities, the inclusion of commodity prices indexes, such as oil prices, as independent variables should mitigate some of the noise generated by these volatile components.

⁴The nominal effective exchange rate is defined as domestic currency units per unit of foreign currencies, which implies that an increase represents a depreciation for home country.

3 Empirical tests

3.1 Panel unit root tests

Before testing for a cointegrating relationship, we investigate panel non-stationarity of the variables included in equation (9). Previous ERPT studies, such as BARHOUMI (2005, 2006) and HOLMES (2006, 2008), has employed first generation panel unit root tests whose main limit is the assumption of cross-sectional independence across units. More specifically, they used the t -bar test proposed by IM et al. (2003), which tests the null hypothesis of non stationarity. This test allows for residual serial correlation and heterogeneity of the dynamics and error variances across groups. The t -bar statistic constructed as a mean of individual Augmented Dickey-Fuller (ADF) statistics and is designed to test the null that all individual units have unit roots. However, it is very important to consider for the possibility of cross-sectional correlation and spillovers amongst countries. The hypothesis of no dependence across the panel members seems to be unrealistic regarding the economic and financial linkages prevailing in the OECD countries.⁵ To deal with this issue, we propose to use the cross-sectionally augmented IPS (CIPS) test as developed by PESARAN (2007). The author suggests a test for cross-sectional dependence and offers a way of eliminating it by augmenting the usual ADF regression with lagged cross-sectional mean and its first-difference to capture the cross-sectional dependence that arises through a single-factor model.⁶

Panel unit root tests results are shown in Table 1 for both levels and first differences variables. Two specifications of the deterministic are considered, i.e. with an intercept only and with an intercept and a linear trend. The critical values for the test statistics based on stochastic simulations are provided in PESARAN (2007). We use the Akaike Information Criterion (AIC) to choose the appropriate lag-length. In the level case, we are unable to reject the null hypothesis that all series are non-stationary in favor of the alternative hypothesis that at least one series from the panel is stationary. For tests on the first differences, we can see that the null of non-stationarity is strongly rejected for all variables belonging to the pass-through equation. We thus conclude that all variables are stationary in first difference.⁷

Table 1: PESARAN's (2007) test for panel unit root

Variables	Level		First difference	
	Intercept only	Intercept & trend	Intercept only	Intercept & trend
mp_{it}	-0.044 (0.482)	-0.630 (0.264)	-7.134 (0.000)	-5.630 (0.000)
e_{it}	-0.812 (0.193)	-0.775 (0.219)	-9.986 (0.000)	-9.164 (0.000)
y_{it}	-0.888 (0.187)	1.043 (0.852)	-7.567 (0.000)	-7.537 (0.000)
w_{it}^*	-0.611 (0.270)	-0.956 (0.170)	-5.791 (0.000)	-6.628 (0.000)

Note: p -values for the null hypothesis of non stationarity are reported between parentheses. Also, the empirical statistics can be compared to the critical value from PESARAN (2007) which are -2.15 for specification with an intercept and -2.65 for specification with intercept and linear time trend, at 5% level. Individual lag lengths are based on Akaike Information Criteria (AIC).

⁵We formally test whether the cross-sectional units are independent by using the cross-section dependence (CD) test statistic of PESARAN (2007). The null hypothesis of no cross-sectional dependence is strongly rejected for our panel of 27 OECD countries. The detailed results of the CD test are available from the authors upon request.

⁶More details on PESARAN's (2007) panel unit root test are provided in Appendix A.

⁷We can compare the empirical statistics to the critical values given in Table II(b) and Table II(c) of PESARAN (2007) at the 5% level for $N = 30$ and $T = 70$.

3.2 Panel cointegration tests

As discussed in BANERJEE, MARCELLINO, and OSBAT (2004), panel cointegration tests can be largely oversized in the presence of cross-unit long-run relationships. Not accounting for such relationships makes it more likely to obtain a finding in favor of cointegration, which may be false. An alternative panel cointegration test was proposed by WESTERLUND (2007). The tests are general enough to allow for a large degree of heterogeneity, both in the long-run cointegrating relationship and in the short-run dynamics, and dependence within as well as across the cross-sectional units. Also, WESTERLUND's (2007) tests have good small-sample properties with small size distortions and high power relative to other popular residual-based panel cointegration tests, such as PEDRONI (1999, 2004).

WESTERLUND (2007) developed four error-correction-based panel cointegration tests. Two tests are designed to test the alternative hypothesis that the panel is cointegrated as a whole, while the other two tests the alternative that at least one unit is cointegrated. The author considers the following error correction model where all variables in levels are assumed to be integrated of order 1:

$$\Delta y_{it} = \delta_i' d_t + \alpha_i (y_{i,t-1} - \beta_i' x_{i,t-1}) + \sum_{j=1}^{p_i} \alpha_{ij} \Delta y_{i,t-j} + \sum_{j=-q_i}^{p_i} \gamma_{ij} \Delta x_{i,t-j} + \varepsilon_{it}, \quad (10)$$

where $d_t(1, t)'$ holds the deterministic components, $\delta' = (\delta_{1i}, \delta_{2i})'$ being the associated vector of parameters. In order to allow for the estimation of the error correction parameter, α_i , by least squares, (10) can be rewritten as

$$\Delta y_{it} = \delta_i' d_t + \alpha_i y_{i,t-1} + \lambda_i' x_{i,t-1} + \sum_{j=1}^{p_i} \alpha_{ij} \Delta y_{i,t-j} + \sum_{j=-q_i}^{p_i} \gamma_{ij} \Delta x_{i,t-j} + \varepsilon_{it} \quad (11)$$

where $\lambda_i' = -\alpha_i \beta_i'$. The parameter α_i corresponds to the speed at which the system corrects back to the long-run equilibrium relationship. WESTERLUND (2007) proposes four tests that are based on least squares estimate of α_i . The tests are designed to test the null hypothesis of no cointegration by testing whether the error correction term in a conditional error correction model is equal to zero. The alternative hypothesis depends on what is being assumed about the homogeneity of α_i .

Two of the four tests are called *group-mean statistics*, do not require the α_i s to be equal, and given as

$$G_\tau = \frac{1}{N} \sum_{j=1}^N \frac{\hat{\alpha}_i}{SE(\hat{\alpha})}, \quad G_\alpha = \frac{1}{N} \sum_{j=1}^N \frac{T \hat{\alpha}_i}{\hat{\alpha}(1)}$$

where $SE(\hat{\alpha})$ is the standard error of $\hat{\alpha}$ and $\hat{\alpha}(1) = \hat{\omega}_{ui} / \hat{\omega}_{yi}$, with $\hat{\omega}_{ui}$ and $\hat{\omega}_{yi}$ are the usual NEWEY and WEST (1994) long-run variance estimators based on $\hat{u}_{it} = \sum_{j=-q_i}^{p_i} \hat{\gamma}_{ij} \Delta x_{i,t-j} + \hat{\varepsilon}_{it}$ and Δy_{it} . The G_τ and G_α statistics test the null hypothesis of no cointegration for all cross-sectional units ($H_0 : \alpha_i = 0$ for all i) against the alternative that there is cointegration for at least one cross-sectional unit ($H_1^G : \alpha_i < 0$ for at least one i). The rejection of null indicates the presence of cointegration for at least one cross-sectional unit in the panel.

The other two tests are called *panel statistics*, assume that α_i is equal for all i , and can be given as follows

$$P_\tau = \frac{\hat{\alpha}_i}{SE(\hat{\alpha})}, \quad P_\alpha = T \hat{\alpha}_i$$

The P_τ and P_α statistics pool information over all the cross-sectional units to test the null of no cointegration for all cross-sectional units ($H_0 : \alpha_i = 0$ for all i) against the alternative of cointegration

for all cross-sectional units ($H_1^P : \alpha_i = \alpha < 0$ for all i). The rejection of null should therefore be taken as the rejection of no cointegration for the panel as a whole.

Table 2 reports WESTERLUND's (2007) panel cointegration test results. We compute both asymptotic and robust bootstrapped p -values. The latter is making inference possible under very general forms of cross-sectional dependence.⁸ According to the group-mean and panel test statistics, we can strongly reject the null of no cointegration. This provides strong evidence of the presence of error correction for individual panel members and for the panel as a whole. We can say that WESTERLUND's (2007) cointegration tests shed light on the existence of a long-run steady-state relationship between the variables entering the ERPT equation.

Table 2: WESTERLUND's (2007) Panel Cointegration Test

Statistic	Value	p -value	Robust p -value
<i>Group-mean statistics</i>			
G_τ	-4.481	0.000	0.000
G_α	-20.086	0.025	0.010
<i>Panel statistics</i>			
P_τ	-11.586	0.001	0.000
P_α	-19.513	0.000	0.000

Note: G_τ and G_α are group mean statistics that test the null of no cointegration for the whole panel against the alternative of cointegration for some countries in the panel. P_τ and P_α are the panel statistics that test the null of no cointegration against the alternative of cointegration for the panel as a whole. Optimal lag and lead lengths are determined by Akaike Information Criterion (AIC). In the last column, we present the bootstrapped p -values which are robust against cross-sectional dependencies. Number of bootstraps is set to 800.

4 Estimation of the long-run ERPT

Once the presence of cointegrating relationship is proved, we can obtain long-run ERPT coefficient by estimating equation 9 in levels. Following PEDRONI (2001), we employ estimation techniques taking into account the heterogeneity of long-run coefficients. Therefore, FMOLS and DOLS Group Mean Estimator can be used to obtain panel data estimates for long-run pass-through. These estimators correct the standard pooled OLS for serial correlation and endogeneity of regressors that are normally present in a long-run relationship.⁹ In our empirical analysis, we emphasize on between-dimension panel estimators. It's worth noting that the between-dimension approach allows for greater flexibility in the presence of heterogeneity across the cointegrating vectors where the pass-through coefficient is allowed to vary.¹⁰ Additionally, the point estimates of the between-dimension estimator can be interpreted as the mean value of the cointegrating vectors, while this is not the case for the within-dimension estimates.¹¹

According to Table 3, long-run pass-through coefficient is statistically significant with the expected positive sign, and the results are fairly robust across estimation techniques. For instance, FM-OLS estimator suggests that one percent depreciation of the nominal exchange rate increases import prices by 0.67%. As we mentioned above, pass-through equation (9) assume unity elasticity of import prices to exchange rate movements in order to account for complete ERPT. However, the null of unity pass-through coefficient ($H_0 : \beta = 1$) is strongly rejected through the different econometric specifications (see t -statistics reported between square brackets in Table 3).

This is an evidence of incomplete ERPT in our sample of 27 OECD countries. On the long run, import prices do not move one-to-one following exchange rate depreciation. These results are in line

⁸Number of replications to obtain bootstrapped p -values is set to 800.

⁹Brief details of these methods are available in Appendix B.

¹⁰Under the within-dimension approach pass-through elasticity would be constrained to be the same value for each country under the alternative hypothesis.

¹¹According to PEDRONI (2001), the between-group FMOLS and DOLS estimators has a much smaller size distortion than the within-group estimators.

with estimates in the literature of exchange rate pass-through into import prices for industrialized countries. For 23 OECD countries, CAMPA and GOLDBERG (2005) find that the average of long run ERPT is 0.64%. In this study, producer-currency pricing (or full pass-through) assumption is rejected for many countries. Using panel cointegration analysis, BARHOUMI (2005, 2006) and HOLMES (2008) reject the pass-through unity for developing countries. In accordance with the conventional wisdom that ERPT is always higher in developing than in developed countries, thus, a partial import prices is more expectable for OECD countries. As a matter of fact, one can think that pass-through would be complete in the long-run due to the gradual full adjustment of prices (as sticky prices tend to be a short run phenomenon).¹² Nevertheless, some microeconomic factors like the pricing strategies of foreign firms can prevent a full adjustment of import prices following currency changes. Exporters of a given product can decide to absorb some of the exchange rate variations within their markups depending on market destination, i.e. by practicing a pricing-to-market strategy.¹³ This would explain why exchange rates are found to be much more volatile empirically than prices, and then pass-through would be incomplete even in the long-run. Also, our findings are in line with the theoretical price discrimination models which allow for a degree of pass-through lower than one even in the long run, as a result of pricing-to-market behaviour (See e.g. CORSETTI, DEDOLA, and LEDUC, 2005).

Table 3: Panel Estimates for 27 OECD countries

Variables	Dependent Variable: Import Price Index	
	Group mean FM-OLS	Group mean DOLS
e_{it}	0.67*** (30.21) [16.71]	0.69*** (26.69) [16.89]
y_{it}	0.27*** (6.15)	0.20*** (6.40)
w_{it}^*	0.68*** (7.09)	0.71*** (6.89)

Note: Group mean FM-OLS and DOLS estimators refer to between-dimension. These estimates include common time dummies. *** indicate statistical significance at the 1 percent level. Pass-through estimates are accompanied by two t -statistics. The t -statistics in parentheses are based on the null of a zero ERPT coefficient ($H_0: \beta = 0$). The t -statistics in square brackets are based on the null of unitary elasticity ($H_0: \beta = 1$).

The evidence of partial ERPT in the long-run would have important policy implications. When imports prices do not fully respond to variation in exchange rate, this would reduce the sensitivity of trade flows to relative price changes, and consequently does not help external adjustment of the economy. For example, a depreciation of the importing country's currency would reduce imports or promote exports less than expected, preventing a prompt adjustment of the trade balance. This outcome represent a key element in understanding the ongoing build-up of the global external imbalances. Moreover, our findings have also important implications for the design of monetary policy and the expectation of inflation. Import prices are a principal channel through which movements in the exchange rate affect domestic prices and hence also the variability of inflation and output. As is well-known, the successful implementation of monetary policy presupposes that central bankers have not only a good understanding of inflation dynamics, but that they are also relatively successful at predicting the future path of inflation. Then, the monetary authorities' forecasts of the future path of inflation must be able to gauge how large are the effects of currency movements. If inflation forecasts are based on estimates that do not take into account for the incomplete degree of ERPT, these forecasts could be overestimating the effects of exchange rate variation on inflation.

¹²See e.g. SMETS and WOUTERS (2002).

¹³Empirically, pricing-to-market can be calculated as the percent change in prices in the exporter's currency due to a one percent change in the exchange rate. Then, the greater the degree of pricing-to-market, the lower the extent of exchange rate pass-through.

5 Conclusion

This paper has examined the long-run exchange rate pass-through (ERPT) into import prices for a panel of 27 OECD countries using panel cointegration approach. Previous empirical panel data studies have neglected the possibility of cross-sectional correlation and spillovers amongst countries. Due to the strong economic and financial linkages between OECD countries, we apply second generation panel unit root and panel cointegration tests which account for possible cross-sectional dependencies among the units included in the panel. We used PESARAN's (2007) panel unit root test to determine the stationarity of panel variables. We find that all variables in levels to be integrated of order 1. Then, we tested whether variables entering the pass-through equation, i.e. import prices, the nominal exchange rate, exporters' costs and demand conditions, are cointegrated. Using WESTERLUND's (2007) tests for panel cointegration, our results suggest the existence of a cointegrating relationship, as implied by the theoretical underpinning. When estimating the long-run pass-through coefficient, both FM-OLS and DOLS estimators show an incomplete pass-through, i.e. import prices sensitivity to exchange rate movements does not exceed 0.70% for our sample of OECD countries. Especially, this outcome would explain the slow adjustment of trade flows to currency changes, even in the long-run, and help us in understanding the ongoing global external imbalances.

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Appendix A. PESARAN’S (2007) test for panel unit root

PESARAN (2007) consider the following simple dynamic linear heterogeneous panel data model:

$$y_{it} = (1 - \rho_i)\mu_i + \rho_i y_{i,t-1} + u_{it}, \quad i = 1, \dots, N; t = 1, \dots, T \quad (12)$$

where the error term u_{it} follow a single common-factor structure

$$u_{it} = \gamma_i f_t + e_{it}, \quad (13)$$

where f_t is an unobserved common factor, γ_i is the corresponding factor loading and e_{it} is an idiosyncratic error term independent across i and independent of the common factor. It is convenient to re-write (12) as

$$\Delta y_{it} = \alpha_i + \beta_i y_{i,t-1} + \gamma_i f_t + e_{it}, \quad (14)$$

where $\alpha_i = (1 - \rho_i)\mu_i$, $\beta_i = -(1 - \rho_i)$ and $\Delta y_{it} = y_{it} - y_{i,t-1}$. The unit root hypothesis of interest, $\rho_i = 1$, can now be expressed as

$$H_0 : \beta_i = 0, \forall i$$

against the possibly heterogeneous alternatives

$$H_1 : \begin{cases} \beta_i < 0 & \text{for } i = 1, 2, \dots, N_1 \\ \beta_i = 0 & \text{for } i = N_1 + 1, N_1 + 2, \dots, N \end{cases} \quad \text{with } 0 < N_1 \leq N$$

To account for the cross-sectional dependence induced by the common factor, PESARAN (2007) suggest to cross-sectionally augment the test equation (14) with cross-sectional averages of the first differences and the lagged levels. The cross-sectionally augmented Dickey-Fuller regression is then given by

$$\Delta y_{it} = a_i + b_i y_{i,t-1} + c_i \bar{y}_{t-1} + d_i \Delta \bar{y}_t + \varepsilon_{it}, \tag{15}$$

where $\bar{y}_{t-1} = \sum_{i=1}^N y_{i,t-1}$, $\Delta \bar{y}_t = \sum_{i=1}^N \Delta y_{it}$ and ε_{it} is the regression error. The individual specific test statistic for the hypothesis $H_{0i} : \beta_i = 0$ for a given i is now the t -statistic of b_i in (15). The statistic is called cross-sectionally augmented Dickey-Fuller ($CADF_i$). The panel unit root for the hypothesis $H_0 : \beta_i = 0$ for all i against the heterogeneous alternative $H_1 : \beta_i < 0$ for some i is given by the cross-sectional average of the $CADF_i$ tests, such that

$$CIPS = \frac{1}{N} \sum_{i=1}^N CADF_i \tag{16}$$

Appendix B. Estimation methods

B1. FM-OLS Mean Group Panel Estimator

We consider the following fixed effect panel cointegrated system:

$$y_{it} = \alpha_i + x'_{it} \beta + \varepsilon_{it}, \quad t = 1, \dots, T, \tag{17}$$

x'_{it} , can in general be a m dimensional vector of regressors which are integrated of order one, that is:

$$x_{it} = +x_{it-1} + u_{it}, \forall i \tag{18}$$

where the vector error process $\xi_{it} = (\varepsilon_{it}, u_{it})'$ is stationary with asymptotic covariance matrix:

$$\Omega_{it} = \lim_{T \rightarrow \infty} E \left[T^{-1} \left(\sum_{t=1}^T \xi_{it} \right) \left(\sum_{t=1}^T \xi'_{it} \right) \right] = \Omega_i^0 + \Gamma_i + \Gamma'_i. \tag{19}$$

Ω_i^0 , is the contemporaneous covariance and, Γ_i , is a weighted sum of autocovariances.

The long run covariance matrix is constructed as follow: $\begin{bmatrix} \Omega_{11i} & \Omega'_{21i} \\ \Omega_{21i} & \Omega_{22i} \end{bmatrix}$, where, Ω_{11i} , is the scalar long run variance of the residual, ε_{it} , and, Ω_{22i} , is the long run covariance among the, u_{it} , and, Ω_{21i} , is vector that gives the long run covariance between the residual, ε_{it} , and each of the u_{it} .

For simplicity, we will refer to, x_{it} , as univariate. So according to PEDRONI (2001), the expression for the group-mean panel FM-OLS estimator (for the between dimension) is given as:

$$\hat{\beta}_{GFM} = N^{-1} \sum_{i=1}^N \left(\sum_{t=1}^T (x_{it} - \bar{x}_i)^2 \right)^{-1} \times \left(\sum_{t=1}^T (x_{it} - \bar{x}_i) y_{it}^* - T \hat{\gamma}_i \right) \tag{20}$$

where $y_{it}^* = (y_{it} - \bar{y}_i) - \frac{\hat{\Omega}_{21i}}{\hat{\Omega}_{22i}} \Delta x_{it}$, and $\hat{\gamma}_i \equiv \hat{\Gamma}_{21i} - \Omega_{21i}^0 - \frac{\hat{\Omega}_{21i}}{\hat{\Omega}_{22i}} (\hat{\Gamma}_{22i} - \Omega_{22i}^0)$, with $y_i = \frac{1}{T} \sum_{t=1}^T y_{it}$ and $x_i = \frac{1}{T} \sum_{t=1}^T x_{it}$ refer to the individual specific means.

The Pedroni between FM-OLS estimator, $\hat{\beta}_{GFM}$, is the average of the FMOLS estimator computed for each individual, $\hat{\beta}_{FM,i}$, that is:

$$\hat{\beta}_{GFM} = N^{-1} \sum_{i=1}^N \hat{\beta}_{FM,i} \tag{21}$$

The associated t -statistic for the between-dimension estimator can be constructed as the average of the t -statistic computed for each individuals of the panel:

$$t_{\hat{\beta}_{GFM}} = N^{-1/2} \sum_{i=1}^N t_{\hat{\beta}_{FM,i}} \tag{22}$$

where $t_{\hat{\beta}_{FM,i}} = \left(\hat{\beta}_{FM,i} - \beta_0 \right) \left(\hat{\Omega}_{11i}^{-1} \sum_{t=1}^T (x_{it} - \bar{x}_i)^2 \right)^{1/2}$.

B2. DOLS Mean Group Panel Estimator

The DOLS regression can be employed by augmenting the cointegrating regression with lead and lagged differences of the regressors to control for endogenous feedback effects. Thus, we can obtain from the following regression:

$$y_{it} = \alpha_i + \beta_i x_{it} + \sum_{k=-K_i}^{K_i} \gamma_{it} \Delta x_{it-k} + \varepsilon_{it}, \tag{23}$$

The group-mean panel DOLS estimator is construct as:

$$\hat{\beta}_{GD} = N^{-1} \sum_{i=1}^N \left(\sum_{t=1}^T Z_{it} Z_{it}' \right)^{-1} \left(\sum_{t=1}^T Z_{it} \tilde{y}_i \right) \tag{24}$$

where $Z_{it} = (x_{it} - \bar{x}_i, \Delta x_{it-K}, \dots, \Delta x_{it-K})$ is a the $2(K+1) \times 1$ vector of regressors and $\tilde{y}_i = y_{it} - \bar{y}_i$.

The DOLS estimator for the i^{th} member of the panel is written as:

$$\hat{\beta}_{D,i} = \left(\sum_{t=1}^T Z_{it} Z_{it}' \right)^{-1} \left(\sum_{t=1}^T Z_{it} \tilde{y}_i \right) \tag{25}$$

So that the between-dimension estimator can be constructed as

$$\hat{\beta}_{GD} = N^{-1} \sum_{i=1}^N \hat{\beta}_{D,i} \tag{26}$$

If the long-run variance of the residuals from the DOLS regression (23) is:

$$\sigma_i^2 = \lim_{T \rightarrow \infty} E \left[T^{-1} \left(\sum_{t=1}^T \varepsilon_{it} \right)^2 \right] \quad (27)$$

According to Pedroni, the associated t-statistic for the between-dimension estimator can be constructed

as:

$$t_{\hat{\beta}_{GD}} = N^{-1/2} \sum_{i=1}^N t_{\hat{\beta}_{D,i}} \quad (28)$$

where $t_{\hat{\beta}_{D,i}} = \left(\hat{\beta}_{D,i} - \beta_0 \right) \left(\hat{\sigma}_i^{-2} \sum_{t=1}^T (x_{it} - \bar{x}_i)^2 \right)^{1/2}$.