

Measurement Matters For Modeling U.S. Import Prices

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1 Introduction

Just how much the dollar should depreciate to restore external balance is the subject of an on-going debate in which little agreement exists. The disagreements stem from several sources but an important one is the responsiveness of U.S. import prices to changes in the value of the dollar. Specifically, Goldberg and Knetter (1997), and Campa and Goldberg (2005) report that half of the movements of the dollar are passed through to import prices whereas Olivei (2002) and Marazzi et al. (2005) find that this pass-through, while significant in the past, has virtually disappeared. Using a new measure of foreign prices, we find that pass-through in the long-run – at the ten-year horizon – has been stable at one-half whereas pass-through in the short-run – at the two-year horizon – has declined markedly over the past decade. We believe these findings contribute to resolving the dispersion of recent estimates.

The general framework to model import prices used by previous studies is given in equation (1):

$$P_m = f(P_y, P_f, P_{com}), \quad (1)$$

where P_m is the price of U.S. imports; P_y is the price of U.S. products; P_f is the foreign price expressed in dollars; and P_{com} is the price of internationally traded commodities. Equation (1) allows for the possibility that foreign exporters seek to retain some parity with general U.S. prices by adjusting their export prices to changes in the prices prevailing in the U.S. market. Such a strategy, known as pricing to market, implies that $\frac{\partial P_m}{\partial P_y} \equiv \gamma > 0$. Equation (1) also allows for the possibility that an increase in the value of the dollar, which raises P_f , is then passed through to P_m . Such a response is known as the exchange-rate pass-through and it implies that $\frac{\partial P_m}{\partial P_f} \equiv \beta > 0$.¹

Equation (1) offers one way to rationalize the dispersion of estimates of β reported in the literature: empirical studies differ in their choice of price measures, econometric formulations, sample periods, and estimation techniques. Another explanation is that this dispersion is reflecting a structural change in the behavior of exporters to the United States that is not adequately captured by recent studies. Indeed, one recent development that qualifies as a structural change is the increased importance of emerging economies as a source of U.S. imports.² To grasp how this possibility shapes our strategy involves a brief expositional detour to which we now turn.

There are two salient features of emerging economies that may affect the behavior of U.S. import prices. First, emerging economies have low production costs relative to developed economies. Second,

¹Not all studies are willing to extend the benefit of the doubt to equation (1). Indeed, recent studies recognize that strategic interactions among producers means that β and γ are not parameters to be estimated but rather endogenous outcomes resulting from demand-supply interactions at the micro level. See Gron and Swenson (2000), Gross and Schmitt (2000), Hellerstein (2004); Bodnar et al. (2002); Gust et al. (2006).

²Recent data show that China displaced Canada as the largest provider of non-fuel goods to the United States.

emerging economies have been increasing their capacity to produce exportable goods at a rapid rate. To the extent that emerging-market producers are pricing their exports to achieve a market share that exhausts their growing production capacity, we would expect two distinct effects on import prices. The first effect is straight-forward: we would expect some downward pressure on import prices as emerging-market exporters price somewhat below prevailing prices in order to gain market share. We would expect this expansion in supply to both increase the total volume of imports and to displace some imports from higher cost producers who are now no longer able to price above their variable costs. The second effect is more subtle and involves the responsiveness of import prices to changes in exchange rates. If the emerging market producers face short-run capacity constraints, then as their share in U.S. imports increases, we would expect import prices to become less responsive to exchange rate movements and more responsive to U.S. price developments.

To understand this change in price responsiveness, it is helpful to focus on the import-supply schedule. To the extent that capacity constraints are binding, the short-run supply schedule becomes more nearly vertical. Thus movements in exchange rates or foreign costs, which induce vertical shifts in the supply schedule, will have a relatively small impact on prices while a shift in the demand schedule will have a relatively large impact on prices. This situation does not imply that foreign prices, or exchange rates, are becoming less important in the determination of U.S. import prices but, rather, that the dynamics of the process are changing. In particular, the long-run (secular) movements are still determined in large part by foreign capacity and costs (with some rightward drift in the supply schedule), but the short-run movements are more closely tied to U.S. demand and U.S. domestic price developments (through the steepening of the supply schedule).

To model this phenomenon econometrically requires a measure of foreign prices that explicitly captures both the growing capacity of emerging-market producers and the fact that the cost structure of these producers is well below that of developed-market producers. Section 2 develops such a measure of foreign prices and contrasts it to existing measures of aggregate foreign prices. Section 2 also looks at the domestic price measures and notes that different measures of U.S. prices carry different implications for modeling import prices. If one chooses the U.S. GDP deflator, then foreign products are assumed to compete with U.S. goods *and* services whereas if one uses the GDP deflator for goods, then foreign products are assumed to compete with U.S. goods.³

Section 3 describes several different econometric specifications that have been used to estimate the parameters of equation (1) and notes how these specifications differ in their characterization of long-

³As long as U.S. services and U.S. goods are not perfect substitute for each other, the measure of U.S. prices matters for estimation.

run behavior. Section 4, using annual observations from 1974 to 2005, estimates several versions of equation (1) and reports on the sensitivity of the estimation results to changes in measures of U.S. and foreign prices, dynamic specifications, and sample periods. The section also estimates the parameters using sample-selection techniques to exclude observations when emerging economies were not important in world trade. Accordingly, Section 4 divides the discussion in two parts: Full-sample Estimation and Sample-selection Estimation. We find that changes in estimation design give rise to a dispersion of estimates, as found in the literature. We also find, using our new measure of foreign prices and recent observations, that the short-run exchange rate pass-through is close to zero but that the long-run pass-through is about one-half. We interpret this result as suggesting that the growing importance of emerging economies does not imply that foreign prices, or exchange rates, are becoming less important in the determination of U.S. import prices but, rather, that the dynamics of the process are changing.

2 Measurement Matters

We measure the price of U.S. imports, P_m , as the multilateral price deflator of U.S. merchandise imports excluding oil, computers, and semiconductors. For work using disaggregate import prices see Yang (1997), Olivei (2002), Campa and Goldberg (2005). For the price of raw materials, P_{com} , we use the IMF's non-fuel commodity price index. For foreign and U.S. domestic prices, however, we consider several alternatives which are discussed below.

2.1 Foreign Prices (P_f)

Measuring foreign prices is of interest because they serve as proxies for foreign costs for which data are not readily available. The general practice for measuring foreign prices aggregates foreign CPIs, expressed in U.S. dollars, using bilateral trade weights. We depart from this practice in several ways. First, we argue that GDP deflators are a better proxy for domestic costs than CPIs. Second, rather than aggregating foreign GDP deflators expressed in U.S. dollars, we construct our measure of P_f so as to ensure consistency with the evolution of U.S. international relative prices. We attain this consistency in two steps. The first step involves measuring U.S. international relative prices, Q :

$$Q_t = \frac{P_{u,t}}{P_{f,t}}, \quad (2)$$

where P_u is the price level of U.S. products. This first step is needed because data for absolute price levels across countries are not easily available whereas the levels of the associated bilateral relative prices

are available. The second step uses equation (2) to solve for P_f taking as given data for Q and P_u .

Implementing these two steps would be straightforward if there were agreement on how to construct Q . As Chinn (2005) notes, however, such an agreement is not currently in place. Indeed, the existing measures of Q constructed by the Federal Reserve, the IMF, and the OECD differ in their mix of countries, in their measure of prices, in their weighting schemes, and in their aggregation methods. The one feature these measures of Q have in common is their reliance on relative price indexes. As discussed further below, such measures cannot fully capture the growing importance of the low-cost producers because indexes indicate, by construction, the percent change of the variable relative to a base year. Our measure focuses on the *level* of international relative prices and not of its change.

2.1.1 Measuring International Relative Prices

Central to our construct of Q is the measurement of the level of the price of U.S. products relative to the level of the price of the i th country, q_{it} . We measure this bilateral relative price as

$$q_{it} = \frac{E_{\$/\$}^i}{PPP_{\$/\$,t}^i},$$

where $E_{\$/\i is the spot dollar exchange rate relative to the i th currency (a rise in $E_{\$/\i constitutes an appreciation of the dollar); and $PPP_{\$/\$,t}^i$ is the PPP exchange rate reported by the Penn World Tables.⁴ This PPP rate is a weighted average of the prices of the i th country relative to U.S. prices using as weights the production levels of the i th country. The prices used by the Penn data are based on highly comparable products across countries and the associated weights account for the full value of GDP. The Penn data are constructed to allow cross-country comparisons and thus well suited to our purposes.

Note that q_{it} differs importantly from the price indexes used in macroeconomics. Typically, these price indexes are constructed to equal 100 in a given year, so it would make no sense to compare the level of price indexes in two countries. In contrast, q_{it} is unitless and easy to interpret: A value of 2 means that the basket of the i th country is twice as expensive in the United States as it is in the i th country.

We aggregate these bilateral relative prices into our construct Q_t using two formulas: a geometric aggregate, denoted as Q_t^g , and a Divisia aggregate, denoted as Q_t^d .⁵ The geometric aggregate is

$$Q_t^g = (q_1)^{w_{1t}} \cdot (q_2)^{w_{2t}} \cdot \dots \cdot (q_n)^{w_{nt}},$$

⁴For an introduction to the Penn World Tables, see Summers and Heston (1991).

⁵The countries we include are the ones included in the Federal Reserve's Broad index of the real effective value of the dollar. Other aggregation rules are available but we use geometric aggregation because of its widespread use in economics. See Bowman (2000).

where w_{it} is the share of the i th country in the aggregate. Several factors affect the aggregate Q_t^g : an increase in U.S. prices raises it, while an increase in foreign prices lowers it; a nominal appreciation of the dollar (an increase in $E_{\$/\text{€}}$) raises Q_t^g . These three factors work through their effect on the bilateral relative prices—the q 's. In addition, changing the weights alone will affect Q_t^g even if none of the underlying prices or exchange rates change. Specifically, emerging-market economies have relatively low production costs and their share in world trade has been increasing. The resulting increase in their share in world markets is captured by the increase in their weights. Loosely speaking, raising the weight on these low-price countries lowers the ‘average’ price abroad and raises U.S. prices relative to the average foreign price, *even if no underlying prices change*.

In contrast, the Divisia aggregate of U.S. international relative prices is a weighted average of growth rates of bilateral relative prices:

$$\frac{Q_t^d}{Q_{t-1}^d} = \prod_{i=1}^N \left(\frac{q_{it}}{q_{i,t-1}} \right)^{w_{it}}, \quad (3)$$

where by convention, $Q_{t=base}^d$ is set equal to 100 to a given period and the index level for all other periods is defined recursively. The chief drawback of this measure for our purposes is that it cannot fully recognize the importance of low-price countries such as China and Mexico. To be sure, the weights for these two countries have increased but their relative prices have been fairly stable which means that the large weights are being multiplied by negligible growth rates. The geometric aggregate does not suffer from this limitation because changes in weights affect the aggregate even if all the relative prices are fixed.

In constructing these aggregates, we consider two types of weights: U.S. bilateral non-oil imports and a broad measure of U.S. trade competitiveness.⁶ As shown in the top panel of figure 1, unlike the choice of weights, the choice of aggregation method matters for measuring the international relative price of U.S. products. Specifically, the geometric aggregate has an upward trend meaning that the price of U.S. products relative to foreign products has been increasing over time. Such a trend owes primarily to shifts in country weights away from the relatively high-price industrial countries toward the lower-price developing economies. In contrast, the Divisia aggregate of relative prices embodies a downward trend in U.S. international relative prices. Accounting for the difference in trends involves recognizing that the Divisia aggregate focuses on the movements in the bilateral real exchange rates alone while ignoring the general shift towards the relatively low-price producers. Capturing this general shift requires aggregation of the levels of bilateral relative prices, which is what geometric aggregation allows.⁷

⁶These competitive weights are those used in the Broad index of the foreign exchange value of the dollar. See Loretan (2005) for the measurement of these weights.

⁷Note, however, that just applying geometric aggregation to the levels of price indexes will not capture the shift towards low-price producers: one would be aggregating growth rates relative to a base period instead of the levels of the relative

2.1.2 Estimating Foreign Prices in Dollars

Given data for U.S. domestic prices—the GDP deflator—and for the aggregate of U.S. international relative prices, we use equation (2) to construct data for foreign prices. Specifically, the measure of foreign prices in dollars implied by Q_t^g is

$$P_{f,t}^g = \frac{P_{u,t}}{Q_t^g}.$$

The measure of foreign prices in dollars implied by Q_t^d is

$$P_{f,t}^d = \frac{P_{u,t}}{Q_t^d}.$$

The bottom panel of figure (1) shows that, given the data for P_u , differences in the evolution of aggregate relative prices translate into differences for the two measures of foreign prices. Specifically, after moving in lock step from 1972 to 1986, the series diverge with P_f^d growing faster than P_f^g . The growing gap between these two series reflects the growing gap associated with the choice of aggregation method.

2.2 U.S. Prices (P_y)

To measure the prices of U.S. products that compete with imports we consider three measures: the Producer Price Index, the overall GDP deflator, and the GDP deflator for goods alone. The distinction between goods and services might matter if one argues that imports of goods compete with U.S. goods, and not with U.S. goods *and* services. That such a distinction is potentially relevant is clear from figure 2. Indeed, since 1970, the price deflator for U.S. goods and services has grown faster than the PPI and the GDP deflator for goods. Furthermore, since 1995, the growth rate of price deflator for goods has a narrow range of fluctuations around zero whereas the growth rate for the overall GDP deflator is positive throughout the sample. For the PPI, the growth rate also fluctuates around zero but the range of the fluctuations exceed that of the deflator for GDP goods.

2.3 Measuring and Modeling

The left panels of figure 3 reveal a close relation between P_f^g and the price of imports. The right panels reveal, in contrast, a growing distancing between P_f^d and the price of imports after 1985. Such a gap can be construed as suggesting a decline in the responsiveness of U.S. import prices to the exchange rate. The data also reveal a growing gap between the price deflator for imports and the price deflator for U.S. goods and services. That distancing is not present if one uses the U.S. deflator for goods alone.

prices.

To evaluate the time-series properties of these variables, Table 1 shows the results from the Augmented Dickey-Fuller tests of stationarity. With one exception, the ADF tests cannot reject a unit-root for these series; the exception is the commodity price index for which the ADF test statistic exceeds by a good margin the 1% critical value. For the price of imports, the ADF test shows a marginal rejection of the unit-root hypothesis at certain lags.

3 Econometric Specifications

For parameter estimation we specify equation (1) as an autoregressive distributed lag:

$$\ln P_{mt} = \alpha + \beta(L) \cdot \ln P_{ft} + \gamma(L) \cdot \ln P_{y,t} + \delta(L) \cdot \ln P_{com,t} + \rho \cdot \ln P_{m,t-1} + u_t, \quad (4)$$

where L is the lag operator and $u_t \sim IN(0, \sigma_u^2)$.⁸ The appeal of equation (4) is its flexibility in differentiating between short-run effects and long run effects. The exchange-rate pass-through is $\beta(1)$ in the short-run and $\frac{\beta(1)}{1-\rho}$ in the long-run. The remaining long-run coefficients are $\frac{\gamma(1)}{1-\rho}$ for the pricing to market and $\frac{\delta(1)}{1-\rho}$ for the price of commodities. The difference between short- and long-run effects is determined by the parameter ρ which measures the persistence of import prices.

A popular alternative to equation (4) explains the growth rate of import prices in terms of growth rates of its determinants:

$$\frac{\Delta P_{mt}}{P_{m,t-1}} = \alpha' + \beta'(L) \cdot \frac{\Delta P_{ft}}{P_{f,t-1}} + \gamma'(L) \cdot \frac{\Delta P_{yt}}{P_{y,t-1}} + \delta'(L) \cdot \frac{\Delta P_{com,t}}{P_{com,t-1}} + \rho' \cdot \frac{\Delta P_{m,t-1}}{P_{m,t-2}} + v_t, \quad (5)$$

where $v_t \sim IN(0, \sigma_v^2)$ and we attach an “'” to the coefficients to emphasize that they differ from those in equation (4). Again, the exchange-rate pass-through is $\beta'(1)$ in the short run and $\frac{\beta'(1)}{1-\rho'}$ in the long-run. The remaining long-run coefficients are $\frac{\gamma'(1)}{1-\rho'}$ for the pricing to market and $\frac{\delta'(1)}{1-\rho'}$ for the price of commodities.

Equations (4) and (5) differ in their characterization of long-run behavior. Specifically, the long-run import price implied by equation (5) is arbitrary because it depends on the value of the initial condition; the long-run import price implied by equation (4) is, however, uniquely determined. Specifically, equation (4) implies that, regardless of the initial condition, the long-run level of import prices is determined as

$$\ln P_m = \frac{\alpha}{1-\rho} + \beta \cdot \ln P_f + \gamma \cdot \ln P_y + \delta \cdot \ln P_{com}, \quad (6)$$

⁸For papers examining the role of non-linearities, see Bussiere (2004); and Pollard and C. Coughlin (2004).

where $\beta = \frac{\beta(1)}{1-\rho}$, $\gamma = \frac{\gamma(1)}{1-\rho}$, and $\delta = \frac{\delta(1)}{1-\rho}$. For these coefficients to be consistent with the predictions of theory, they need to meet the sign restrictions ($\beta > 0$, $\gamma > 0$, $\delta > 0$) and the homogeneity condition ($\beta + \gamma + \delta = 1$); the long-run coefficients for equation (5) need to meet comparable conditions.⁹

4 Econometric Results

4.1 Full-sample Estimation

Parameter estimation rests on the General-to-Specific strategy. The general formulations correspond to equations (4) and (5) where we include current values and 1 lag for each predetermined variable; the specific formulations are obtained by implementing the automated-specification algorithm developed by Hendry and Krolzig (2001). Their algorithm combines least squares with a selection criteria that excludes formulations lacking parameter constancy, white-noise residuals, and insignificant variables. Further, the critical values used in model selection are not fixed in advance but, rather, are calculated sequentially to recognize the joint nature of model specification and parameter estimation.

Table 2 reports the long-run coefficient estimates for the specific formulation associated with equation (4) along with the evaluation of the properties of the residuals; Table 3 reports the associated results for equation (5). To facilitate the discussion, figures 4-6 show the sensitivity of these estimates to changes in

1. the econometric specification: equation (4) versus equation (5);
2. the measure of U.S. prices, P_{yt} : GDP deflator for goods and services (G&S), GDP deflator for goods (Goods), Producer Price Index (PPI);
3. the measure of foreign price, P_{ft} : geometric aggregate and Divisia aggregate;
4. the weighting scheme, w_{it} : U.S. bilateral non-oil imports (non-oil) to using U.S. trade competitiveness (broad);
5. the estimation period: 1974-2000 and 1974-2005.

Based on the full sample, figure 4 reports the long-run estimates of the pricing to market coefficient, $\hat{\gamma}$, and of exchange-rate pass-through coefficient, $\hat{\beta}$. Inspection of the results reveals several findings. First, as shown in the top panel, $\hat{\gamma}$ is positive and significant except when the estimation uses the formulation in levels with the Divisia measure of foreign prices along with either the U.S. PPI or the U.S. overall

⁹Note that the inclusion of an intercept in equation (5) implies that the implied level of the price of imports will include a trend. Thus P_m changes at a rate of $\frac{\alpha'}{1-\rho}$ even if all other prices are constant.

GDP deflator. Second, as shown in the bottom panel, $\hat{\beta}$ is positive, significantly greater than zero, and exceeds 1/3. Third, values of $\hat{\beta}$ based on the equation in levels, equation (4), with the geometric foreign price are close to 0.5 and are, in general, larger than estimates based on the Divisia foreign price.

The role of commodity prices depends on the specification. For the equation in levels (Table 2) commodity prices are not relevant whereas for the equation in growth rates (Table 3), they are. Such a difference is not surprising because, as Table 1 shows, the level of the price of commodities is stationary. In terms of the estimate of persistence, $\hat{\rho}$, the results for the equation in levels (Table 2) indicate that $\hat{\rho}$ is significant and well below one; for the equation in growth rates (Table 3), $\hat{\rho}$ is not statistically different from zero. In terms of the properties of the residuals, the evidence cannot reject the hypotheses of serial independence and homoskedasticity for either specification.

Finding that the smallest value of $\hat{\beta}$ exceeds 1/3 is at odds with studies that focus on recent observations alone and report values of $\hat{\beta}$ close to zero. One possibility is that recent observations are responsible for a decline in $\hat{\beta}$. In such a case, one should expect that excluding them from our sample would raise the value of $\hat{\beta}$. To explore this possibility, we re-estimate the coefficients with a sample that ends in 2000 and compare the resulting estimates to those based on the full sample.

Figure 5 shows how the value of $\hat{\beta}$ changes in response to alternative sample periods while controlling for price measures and functional forms.¹⁰ We find no evidence of a higher long-run exchange-rate pass-through for the sample ending in 2000. Instead, for the level specifications and some U.S. price measures, we find that extending the sample *increases* noticeably the value of $\hat{\beta}$. As shown in figure 6, the estimated pricing-to-market coefficient, $\hat{\gamma}$, is somewhat more sensitive to sample period, especially for the level specifications based on the Divisia measure of foreign prices.

Several conclusions emerge from these estimation results. First, the long-run effect of exchange rates on import prices is at least one-third across all regressions. Second, the mere inclusion of recent observations does not significantly lower the value of $\hat{\beta}$. Third, the measurement of foreign prices matters for estimating the exchange-rate pass-through. Specifically, values of $\hat{\beta}$ based on the geometric foreign price are larger (close to 0.5) and less sensitive to changes in specification than estimates based on the Divisia foreign price. Finally, changes in estimation design give rise to a dispersion of estimates, as found in the literature. Such a dispersion is problematic for policy applications: seemingly minor changes in estimation design carry radically different implications regarding the strength of the expenditure-switching effect of a change in the exchange value of the dollar.

To discriminate among these specifications, we argue that policy applications benefit from formulations

¹⁰The General to Specific strategy is applied to each sample period separately. As a result, the specific formulation for the shorter sample is not influenced by observations outside the sample period.

that exhibit three properties: constancy of long-run parameters, white-noise residuals, and consistency with price homogeneity. Applying this criteria yields two specifications, reproduced in Table 4. One specification uses the formulation in levels along with the geometric foreign price and the U.S. deflator for GDP goods; this specification is labeled as S^g . The second specification uses the formulation in growth rates along with the Broad weighted Divisia measure of foreign price and the U.S. PPI; this specification is labeled as S^d .¹¹

These two specifications have similar values for $\hat{\beta}$: 0.5 for S^g and 0.4 for S^d . However, these specifications differ in their implications for the long-run: the coefficient for S^g refers to a long-run response of the *level* of import prices whereas the coefficient for S^d refers to the short run (one year) response of the *growth rate* of import prices. This distinction would not be relevant if the estimated adjustment delay in S^g were brief, but it is not. Figure 7 shows how import prices respond to a hypothetical one-percent sustained depreciation of the dollar.¹² The results reveal several features of interest. First, the models differ greatly in their short-run responses. Specifically, after one year, S^d predicts that import prices will increase by 0.4 percent (blue line) whereas S^g shows a negligible response (black line).¹³ After four years, however, the two specifications predict roughly the same effect on import prices. Second, to emphasize how changes in the estimation sample have changed the character of dynamic adjustment, the figure also includes the response from S^g using parameter estimates based on data through 2000 (red line). The responses from these two sample periods converge to virtually identical values in the long run. Their short-run responses are, however, quite different: the specification based on data through 2000 has a higher short-run response and a quicker adjustment to the long-run than the estimate based on data through 2005. We interpret this finding as being consistent with the idea that the increased importance of emerging economies has changed the dynamics of the process of import-price determination.

Beyond stating the obvious—namely, that there is an important difference between short and long run effects, these results are not powerful enough to discriminate between S^g and S^d . To that end, we examine the out-of-sample accuracy of these two formulations. Specifically, we apply the General-to-Specific algorithm with data through 2000 and then use dynamic simulations to generate ex-post forecasts

¹¹Table 4 also offers an informal assessment of the degree to which there is cointegration among prices. For the geometric measure of foreign price, the best equation (parameter constancy, white-noise residuals, homogeneity) explains the level of the import prices. The cointegration test for this equation involves comparing the t-statistic of one minus the coefficient for the lagged dependent variable against the Ericsson-Mackinnon critical value. This coefficient is 0.42 (=1-0.58) and its standard error is 0.09; the associated t-statistic is 4.67. The Ericsson-Mackinnon critical value (5%) is 3.33 (Ericsson, 2006, p.302, table 6.7). Thus the evidence cannot reject cointegration if one uses the geometric measure of foreign prices. For the Divisia measure of foreign price, the best equation explains the growth rate of import prices. As a result, the evidence does not support cointegration in this case.

¹²The response is computed as $(\beta_0 + \beta_1 \cdot L) \cdot (1 + \rho \cdot L + \rho^2 \cdot L^2 + \dots)$.

¹³The estimates for S^g show a negligible response in short-run pass-through coefficients because β_0 and β_1 have opposite signs with $\beta_0 + \beta_1 > 0$.

for 2001-2005.¹⁴ Figure 8 shows the 95 percent confidence bands of these ex-post forecasts.¹⁵ The results indicate that, over the first two years, both equations have statistically insignificant prediction errors but that S^d is much more accurate than S^g . Thus, if the sole purpose of estimation is short-term forecasting, then S^d is the preferred specification. The evidence for horizons beyond two years is, however, less supportive of S^d . Even though both equations show statistically significant prediction errors, the root-mean squared error over the five-year horizon is six percent for S^g but eight percent for S^d . Thus, if the policy application extends beyond two years then S^g carries greater precision.

At this point we would conclude that, of the two formulations that exhibit white-noise residuals and consistency with the homogeneity postulate, the one with the lowest average forecast error in the medium term uses our geometric measure of foreign prices. It has a long-run pass-through of one-half. This formulation captures the changing nature of the import-price process by using a measure of foreign prices that reflects the growing importance of emerging economies. There is, however, an alternative strategy that retains the conventional measures of foreign price and uses sample-selection methods so as to exclude observations when emerging economies were not important in world trade. In the next subsection we weigh the choice of retaining current (Divisia based) measures of foreign prices and shortening the sample period against replacing the Divisia measures of foreign prices with the geometric alternative and retaining the full sample period.

4.2 Sample-selection Estimation

The result of the recent decline in the pass-through is based on quarterly data whereas our data are annual. Thus we begin by confirming that this same feature is also present in the annual analogue to these models. To this end, we apply rolling regressions to the growth equation (5) using the Divisia measure of P_{ft} and the producer price index for $P_{y,t}$ and a window of 15 years starting in 1974.¹⁶ Figure 9 shows the 95 percent confidence interval for the estimated pass-through coefficient of equation (5), $\widehat{\beta}'$. The results reveal a sustained decline in $\widehat{\beta}'$ from 0.45 in 1995 to 0.17 in 2005, a pattern in line with the findings of Olivei (2002) Marazzi and Sheets (2006).¹⁷

We now ask whether this decline in $\widehat{\beta}'$ is robust to model specification. In choosing an alternative

¹⁴The re-application of the General to Specific algorithm avoids the specifications being influenced by observations beyond 2000.

¹⁵For the equation in growth rates, we generate the prediction for the level by augmenting the model with the definition of the logarithm of the price level: $\ln P_{mt} = \ln P_{m,t-1} + d \ln P_{mt}$.

¹⁶Rolling regressions use a fixed sample the dates of which (the window) change systematically by excluding the earliest date and including the next date.

¹⁷To attain greater comparability with previous results, we also treat the specification as known and invariant from one sample to another: only the estimated parameters change as the sample changes. Operationally, we implement first a General-to-Specific strategy using the full sample and then apply rolling regressions to the selected specification. The alternative would be to find a specific formulation for each estimation window. Studies reporting rolling-regression estimates do not search window by window and so we retain their practice.

formulation, we follow Ericsson (2006, Table 6.2) who shows that equation (4) can be expressed as an error-correction formulation:

$$\begin{aligned} \frac{\Delta P_{mt}}{P_{m,t-1}} = & \alpha' + \beta'(L) \cdot \frac{\Delta P_{ft}}{P_{f,t-1}} + \gamma'(L) \cdot \frac{\Delta P_{yt}}{P_{y,t-1}} + \delta'(L) \cdot \frac{\Delta P_{com,t}}{P_{com,t-1}} + \rho' \cdot \frac{\Delta P_{m,t-1}}{P_{m,t-2}} \\ & + \tilde{\rho} \cdot \ln P_{m,t-1} + \tilde{\beta} \cdot \ln P_{f,t-1} + \tilde{\gamma} \cdot \ln P_{y,t-1} + \tilde{\delta} \cdot \ln P_{com,t-1} + u_t. \end{aligned} \quad (7)$$

Equation (7) is appealing because it differentiates between short- and long-run responses. The coefficients associated with the growth rates of the right-hand side variables represent short-run responses; the coefficients associated with the variables in levels capture the adjustment of the price of imports to its long-run value. Indeed, the resulting expression for the long-run level of import prices is equation (6) above where $\beta = -\frac{\tilde{\beta}}{\rho}$, $\gamma = -\frac{\tilde{\gamma}}{\rho}$, $\delta = -\frac{\tilde{\delta}}{\rho}$, $\tilde{\rho} = (\rho - 1)$, and ρ is the coefficient of the lagged import price in for the equation in levels, equation (4).

For estimation, we implement first a General-to-Specific strategy using the full sample and then apply rolling regressions to the selected specification. The top left panel of figure 10 shows that $\widehat{\beta}$ declines from 0.5 in 1990 to 0.2 in 2005, just as in the rolling-regression results reported in figure 9. This finding suggests that the short-run pass-through has declined regardless of functional form, a finding consistent with the results reported by Olivei (2002) and Marazzi and Sheets (2006). The top right and bottom panels of figure 10 show that both $\widehat{\tilde{\rho}}$ and $\widehat{\tilde{\beta}}$ start at zero but increase (in absolute value) and remain constant since 1995. In particular, figure 11 shows that the implied estimate of the long-run pass-through, $\widehat{\beta} = -\frac{\widehat{\tilde{\beta}}}{\widehat{\tilde{\rho}}}$, has fluctuated between 0.4 and 0.5 with a sample mean of 0.48, which is close the value of $\widehat{\beta}$ of S^g . These findings support our view that the growing importance of emerging economies does not imply that foreign prices, or exchange rates, are becoming less important in the determination of U.S. import prices but, rather, that the dynamics of the process are changing. Indeed, the short-run exchange rate pass-through is now close to zero but the long-run pass-through is about one-half.

Again, this difference between short- and long-run responses would not be relevant if the adjustment process were swift. To estimate the delay in adjusting to the long-run, we examine the response of import prices to a sustained, one-percent increase in the value of the dollar.¹⁸ Figure 12 shows a similarity of short-run responses: 0.17 for the formulation in growth rates (yellow line) and 0.25 for the error-correction model (green line). This similarity is, however, short-lived with the response from the error-correction model being more than twice as large as that from the growth rate equation after three years. Further, figure 12 suggests that the dispersion in estimates found in the literature might be due to a failure to

¹⁸We use parameter estimates based on the last window of the rolling regression (1991-2005).

differentiate between short-run and long-run responses in the level of import prices.

Figure 12 also suggests that one can capture the increased role of emerging economies in two ways: by either modifying methods to measure foreign prices (black line) or by using conventional measures of prices combined with rolling regressions (green line). However, lacking a test to discriminate between the two approaches, we offer two reasons why modeling import prices benefits from a strategy that emphasizes measurement as opposed to sample selection. First, rolling regressions embody a tension relevant to policy work: whereas the parameter estimates are assumed to be constant during the forecast period, they are assumed to change during the estimation period. Second, as Marazzi and Sheets (2006) note, estimates from rolling regressions are quite sensitive to the choice of the window used for estimation. Without objective criteria for choosing the estimation window, this introduces another subjective, but important, element to the modeling process.

5 Conclusions

This paper offers an empirical analysis of the behavior of U.S. import prices with a focus on assessing the dispersion of estimated pass-through of exchange rates. We find that, whereas the pass-through in the short-run has declined, the long-run pass-through has remained stable. We believe that these findings argue for employing formulations that differentiate between short- and long-run responses. We also argue that the one force behind the change in dynamics of the import-price process is the greater presence of producers from emerging economies. And we show that their effect on import prices can be captured with our measure of foreign prices.

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Table 1 : Augmented Dickey-Fuller Tests-1974-2005

Variable	Lags	t-adf	beta Y_1	SER	AIC	Variable	Lags	t-adf	beta Y_1	SER	AIC
Import Price	3	-2.896	0.899	0.030	-6.891	Commodity Price	3	-4.712**	0.292	0.085	-4.783
	2	-3.206*	0.907	0.029	-6.950		2	-4.830**	0.243	0.090	-4.699
	1	-2.173	0.938	0.032	-6.808		1	-4.936**	0.266	0.088	-4.755
	0	-3.646*	0.906	0.034	-6.713		0	-3.952**	0.428	0.096	-4.620
GDP Deflator Goods & Services	3	-2.014	0.976	0.009	-9.321	GDP Deflator Goods	3	-2.276	0.946	0.012	-8.756
	2	-2.574	0.971	0.009	-9.338		2	-2.613	0.941	0.011	-8.810
	1	-1.721	0.981	0.009	-9.234		1	-2.434	0.950	0.011	-8.841
	0	-8.198**	0.940	0.012	-8.752		0	-8.198**	0.891	0.013	-8.556
Foreign Price: Geometric Non-oil Weights	3	-2.410	0.901	0.047	-5.954	Foreign Price: Geometric Trade Weights	3	-2.302	0.904	0.050	-5.857
	2	-2.281	0.912	0.047	-5.994		2	-2.152	0.915	0.049	-5.894
	1	-2.214	0.917	0.046	-6.044		1	-2.080	0.920	0.049	-5.942
	0	-2.971*	0.882	0.052	-5.838		0	-2.760	0.887	0.055	-5.747
Foreign Price: Divisia Non-oil Weights	3	-1.790	0.947	0.049	-5.866	Foreign Price: Divisia Trade Weights	3	-1.860	0.944	0.050	-5.845
	2	-1.723	0.953	0.049	-5.920		2	-1.711	0.951	0.050	-5.888
	1	-1.572	0.958	0.049	-5.955		1	-1.547	0.957	0.050	-5.917
	0	-2.137	0.937	0.055	-5.741		0	-2.108	0.935	0.056	-5.692
Producer Price Index	3	-2.238	0.942	0.027	-7.051						
	2	-3.118*	0.931	0.027	-7.089						
	1	-2.036	0.949	0.032	-6.823						
	0	-4.723**	0.902	0.035	-6.637						

Legend

t-ADF: augmented Dickey-Fuller Test; an'*' means that the test rejects the hypothesis of a unit root

beta Y-1: Coefficient of the level of the lagged dependent variable

AIC: Akaike information criterion

Rejection values 5%=-2.96; 1%=-3.67. Constant included.

Table 2: Long-run Coefficients: Level Specification--Sensitivity to Foreign Prices, U.S. Prices, Weighting Schemes, and Sample Periods

Foreign Price	U.S. Price	Weights	Sample	Pricing to Market		Pass-through		Commodity		Lagged Dep. Var.		Intercept		Homogeneity		Residuals (a)			
				coeff.	se	coeff.	se	coeff.	se	coeff.	se	coeff.	se	coeff.	se	SER	Independ	Homosk	
Geometric																			
	Goods & Services	Nonoil	1974-2000	0.72	0.04	0.26	0.03	0.10	0.03	0e	--	-0.31	0.12	1.09	0.06	0.014	0.35	0.54	
				1974-2005	0.49	0.05	0.46	0.04	0e	--	0.29	0.06	0.36	0.13	0.95	0.06	0.018	0.02	0.76
		Broad	1974-2000	0.71	0.04	0.27	0.03	0.10	0.02	0e	--	-0.27	0.11	1.08	0.05	0.013	0.36	0.36	
				1974-2005	0.47	0.05	0.47	0.04	0e	--	0.29	0.06	0.41	0.13	0.94	0.06	0.017	0.02	0.66
	Goods	Nonoil	1974-2000	0.51	0.07	0.52	0.08	0e	--	0e	--	0e	--	1.03	0.11	0.028	0.07	0.57	
				1974-2005	0.56	0.12	0.47	0.12	0e	--	0.64	0.09	0e	--	1.03	0.17	0.026	0.39	0.69
		Broad	1974-2000	0.50	0.06	0.52	0.06	0e	--	0.33	0.06	0e	--	1.03	0.08	0.025	0.08	0.78	
				1974-2005	0.53	0.09	0.49	0.10	0e	--	0.58	0.09	0e	--	1.03	0.13	0.026	0.18	0.62
	PPI	Nonoil	1974-2000	0.52	0.08	0.48	0.09	0e	--	0.56	0.16	0e	--	1.00	0.12	0.020	0.16	0.76	
				1974-2005	0.35	0.16	0.57	0.12	0e	--	0.66	0.15	0.45	0.35	0.91	0.20	0.020	0.49	0.51
		Broad	1974-2000	0.54	0.05	0.46	0.06	0e	--	0.41	0.16	0e	--	1.00	0.08	0.018	0.12	0.76	
				1974-2005	0.46	0.05	0.50	0.05	0e	--	0.27	0.08	0.23	0.13	0.95	0.07	0.020	0.05	0.52
Divisia																			
	Goods & Services	Nonoil	1974-2000	0.77	0.07	0.18	0.05	0.13	0.04	0.33	0.08	-0.42	0.20	1.08	0.09	0.014	0.44	0.68	
				1974-2005	0.26	0.21	0.48	0.18	0e	--	0.69	0.08	0.98	0.40	0.75	0.28	0.021	0.02	0.09
		Broad	1974-2000	0.78	0.07	0.17	0.05	0.13	0.04	0.32	0.08	-0.39	0.19	1.08	0.09	0.014	0.40	0.50	
				1974-2005	0.27	0.20	0.47	0.14	0e	--	0.68	0.08	1.02	0.38	0.74	0.24	0.022	0.16	0.05
	Goods	Nonoil	1974-2000	0.55	0.04	0.42	0.04	0e	--	0.30	0.08	0e	--	0.97	0.06	0.022	0.03	0.53	
				1974-2005	0.73	0.08	0.35	0.04	0.06	0.03	0.18	0.09	-0.73	0.25	1.14	0.09	0.018	0.01	0.96
		Broad	1974-2000	0.79	0.07	0.34	0.03	0e	--	0e	--	-0.70	0.19	1.13	0.08	0.020	0.04	0.88	
				1974-2005	0.78	0.06	0.34	0.03	0e	--	0e	--	-0.70	0.17	1.13	0.07	0.019	0.02	0.79
	PPI	Nonoil	1974-2000	0.63	0.03	0.32	0.03	0e	--	0.33	0.13	0e	--	0.95	0.04	0.015	0.22	0.29	
				1974-2005	0.15	0.28	0.47	0.16	0e	--	0.78	0.12	1.51	0.74	0.63	0.32	0.020	0.46	0.37
		Broad	1974-2000	0.63	0.03	0.32	0.03	0e	--	0.29	0.12	0e	--	0.95	0.04	0.014	0.23	0.39	
				1974-2005	0.19	0.25	0.46	0.14	0e	--	0.76	0.12	1.41	0.64	0.65	0.29	0.020	0.38	0.48

0e: Algorithm excludes explanatory variable from specification.

(a): Entries for "Independ" and "Homosk" are significance levels for rejecting the hypothesis of serial independence and homoskedasticity.

Table 3: Long-run Coefficients; Growth Specification--Sensitivity to Foreign Prices, U.S. Prices, Weighting Schemes, and Sample Periods

Foreign Price	U.S. Price	Weights	Sample	Pricing to Market		Pass-through		Commodity		Lagged Dep. Var		Intercept		Homogeneity		Residuals(a)			
				coeff	se	coeff	se	coeff	se	coeff	se	coeff	se	coeff	se	SER	Independ.	Homosk	
Geometric																			
	Goods & Services	Nonoil	1974-2000	0.73	0.11	0.37	0.09	0.09	0.03	0e	--	0e	--	1.18	0.15	0.022	0.91	0.82	
					1974-2005	0.71	0.11	0.36	0.09	0.07	0.03	0e	--	0e	--	1.14	0.14	0.022	0.96
		Broad	1974-2000	0.75	0.11	0.35	0.08	0.09	0.03	0e	--	0e	--	1.19	0.14	0.022	0.23	0.91	
					1974-2005	0.72	0.11	0.35	0.08	0.08	0.03	0e	--	0e	--	1.15	0.14	0.022	0.89
	Goods	Nonoil	1974-2000	0.83	0.16	0.42	0.10	0.10	0.04	0e	--	0e	--	1.36	0.19	0.026	0.70	0.98	
					1974-2005	0.85	0.15	0.41	0.09	0.09	0.04	0e	--	0e	--	1.36	0.18	0.024	0.76
		Broad	1974-2000	0.86	0.16	0.40	0.10	0.11	0.04	0e	--	0e	--	1.37	0.19	0.026	0.60	0.97	
					1974-2005	0.88	0.14	0.39	0.08	0.10	0.04	0e	--	0e	--	1.37	0.17	0.024	0.62
	PPI	Nonoil	1974-2000	0.58	0.10	0.47	0.08	0.06	0.03	0e	--	0e	--	1.11	0.13	0.022	0.57	0.05	
					1974-2005	0.52	0.11	0.44	0.09	0.08	0.05	0e	--	0e	--	1.04	0.15	0.024	0.57
		Broad	1974-2000	0.60	0.09	0.45	0.07	0.06	0.03	0e	--	0e	--	1.11	0.12	0.022	0.35	0.04	
					1974-2005	0.49	0.12	0.46	0.10	0.11	0.06	0.10	0.11	0e	--	1.05	0.17	0.024	0.53
Divisia																			
	Goods & Services	Nonoil	1974-2000	0.72	0.12	0.33	0.09	0.09	0.04	0e	--	0e	--	1.14	0.15	0.023	0.96	0.55	
					1974-2005	1.01	0.16	0.36	0.08	0.08	0.03	0e	--	-0.02	0.01	1.44	0.18	0.021	0.99
		Broad	1974-2000	0.72	0.12	0.33	0.08	0.10	0.04	0e	--	0e	--	1.14	0.15	0.023	0.94	0.79	
					1974-2005	1.00	0.16	0.34	0.08	0.08	0.03	0e	--	-0.02	0.01	1.43	0.18	0.022	0.96
	Goods	Nonoil	1974-2000	0.83	0.16	0.39	0.09	0.10	0.04	0e	--	0e	--	1.33	0.19	0.026	0.68	0.89	
					1974-2005	0.77	0.15	0.36	0.08	0.15	0.05	0e	--	0e	--	1.28	0.17	0.023	0.49
		Broad	1974-2000	0.83	0.16	0.39	0.09	0.11	0.04	0e	--	0e	--	1.32	0.19	0.026	0.64	0.93	
					1974-2005	0.77	0.15	0.35	0.08	0.16	0.05	0e	--	0e	--	1.28	0.17	0.023	0.40
	PPI	Nonoil	1974-2000	0.55	0.10	0.41	0.07	0.12	0.05	0e	--	0e	--	1.08	0.13	0.022	0.24	0.87	
					1974-2005	0.59	0.12	0.44	0.09	0.08	0.06	0e	--	-0.01	0.01	1.11	0.16	0.024	0.68
		Broad	1974-2000	0.54	0.10	0.40	0.07	0.13	0.05	0e	--	0e	--	1.07	0.13	0.022	0.20	0.72	
					1974-2005	0.58	0.12	0.42	0.09	0.09	0.06	0e	--	-0.01	0.01	1.09	0.16	0.025	0.60

0e: Algorithm excludes explanatory variable from specification.

(a): Entries for "Independ" and "Homosk" are significance levels for rejecting the hypothesis of serial independence and homoskedasticity.

Table 4: Long-run Coefficient Estimates – OLS, 1974-2005

Selected Specifications				
(standard errors in parentheses)				
	Formulations ^a			
	S^g		S^d	
	1974-2000	1974-2005	1974-2000	1974-2005
Pricing to Market	0.50 (0.06)	0.53 (0.09)	0.54 (0.10)	0.58 (0.12)
Pass-through	0.52 (0.06)	0.49 (0.10)	0.40 (0.07)	0.42 (0.09)
Persistence	0.33 (0.06)	0.58 (0.09)	-	-
Homogeneity ^b	1.03 (0.08)	1.03 (0.14)	1.07 (0.13)	1.09 (0.15)

^a Formulations

S^g : Equation (4) with geometric measure of foreign prices and GDP-goods deflator for U.S. prices.

S^d : Equation (5) with Divisia measure of foreign prices and PPI for U.S. prices.

^b Sum of pricing to market and pass-through coefficients; standard error computed without taking into account the covariance of the estimates.

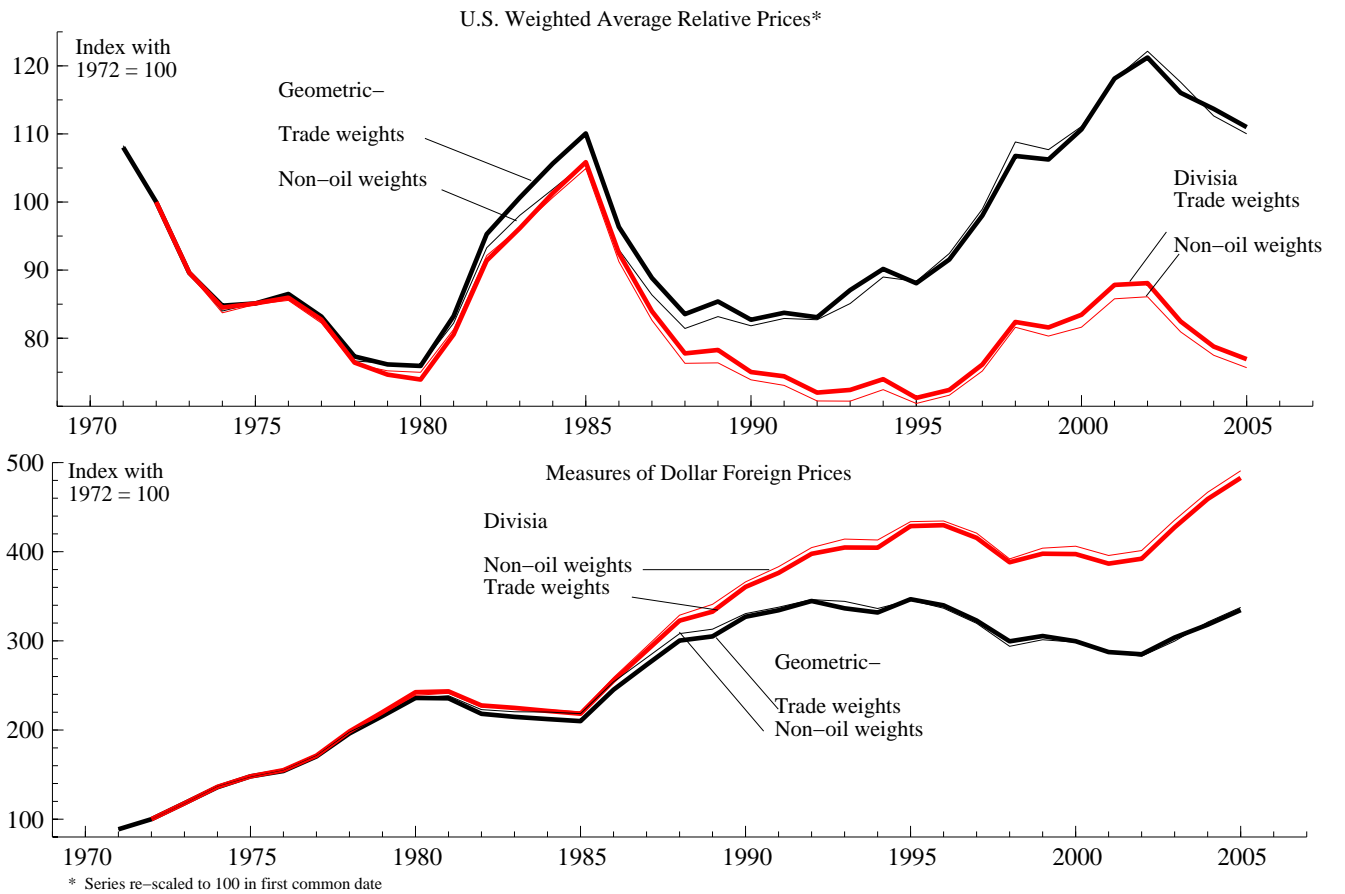


Figure 1: U.S. International Relative Prices and Aggregate Foreign-GDP Deflators

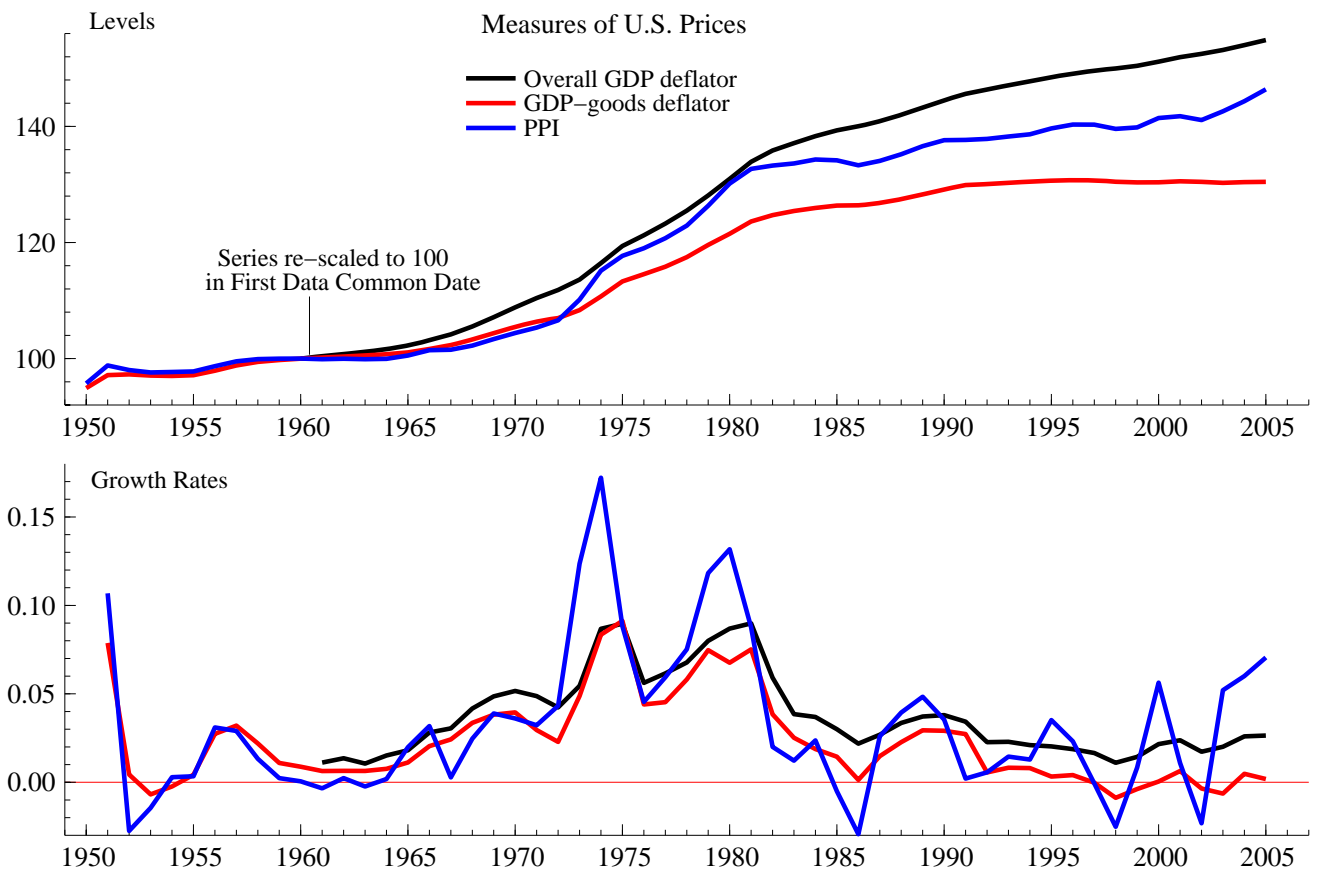


Figure 2: Alternative Measures of U.S. Prices

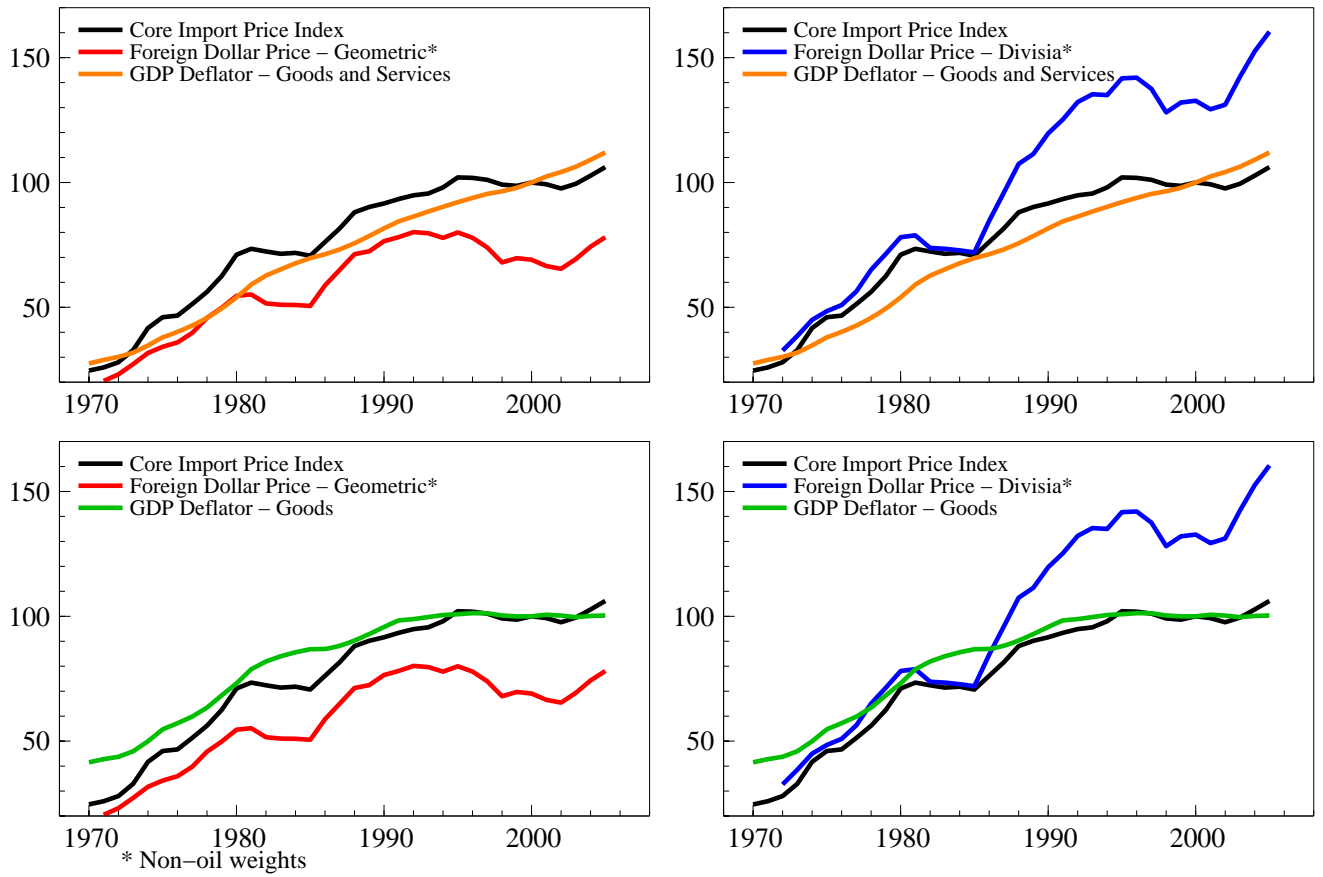


Figure 3: Import Prices, Domestic and Foreign Price Levels

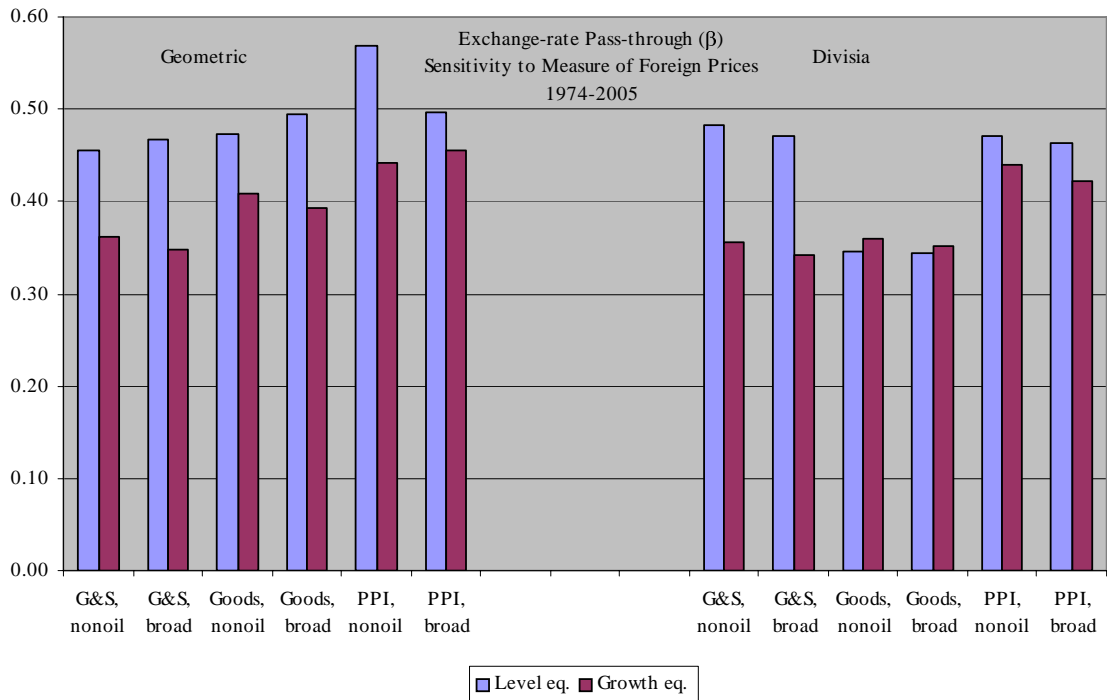
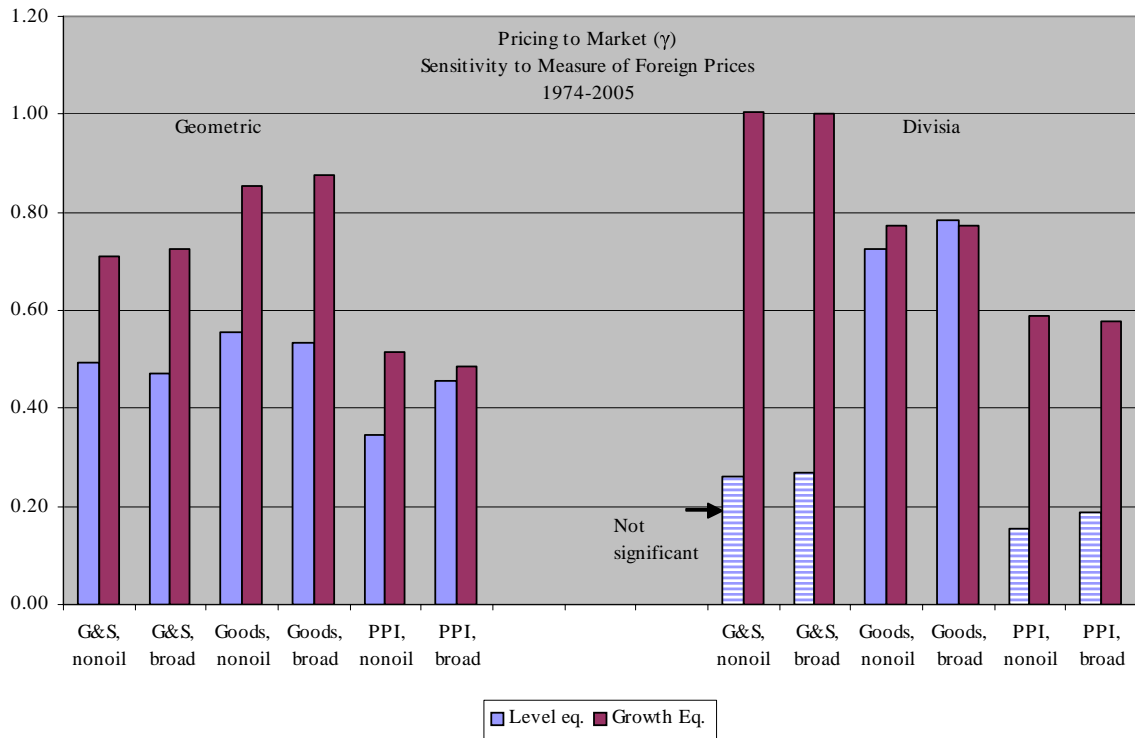


Figure 4: Long-run Coefficient Estimates – Sensitivity to Estimation Design

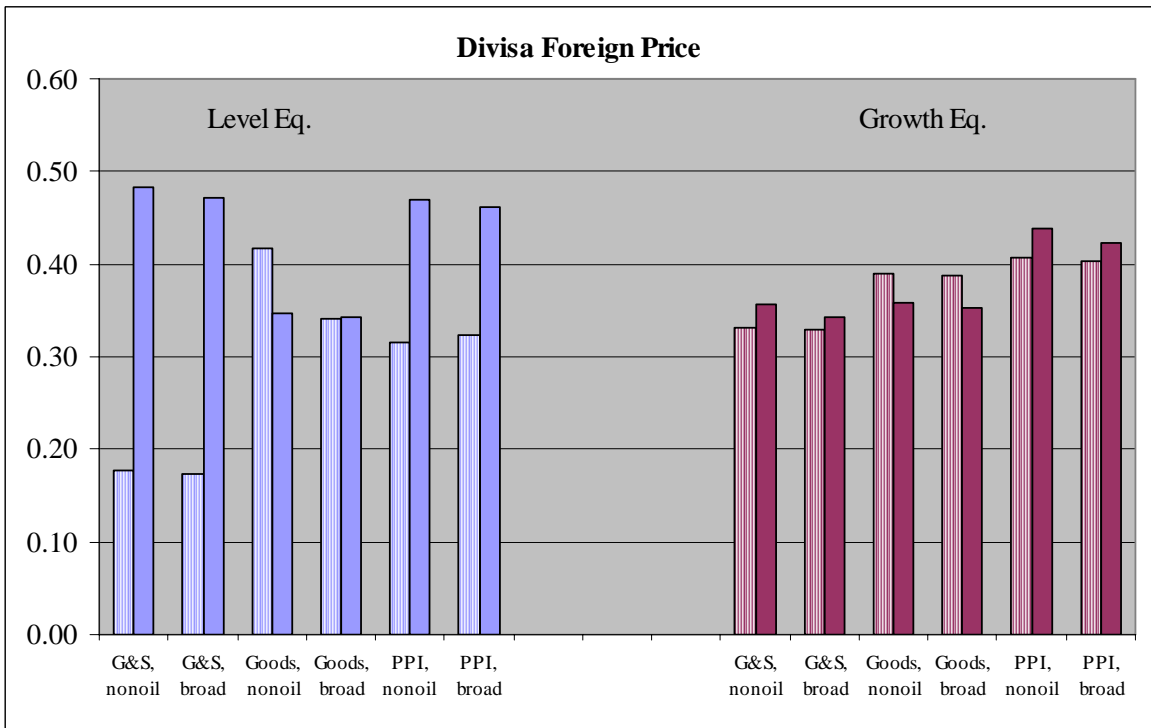
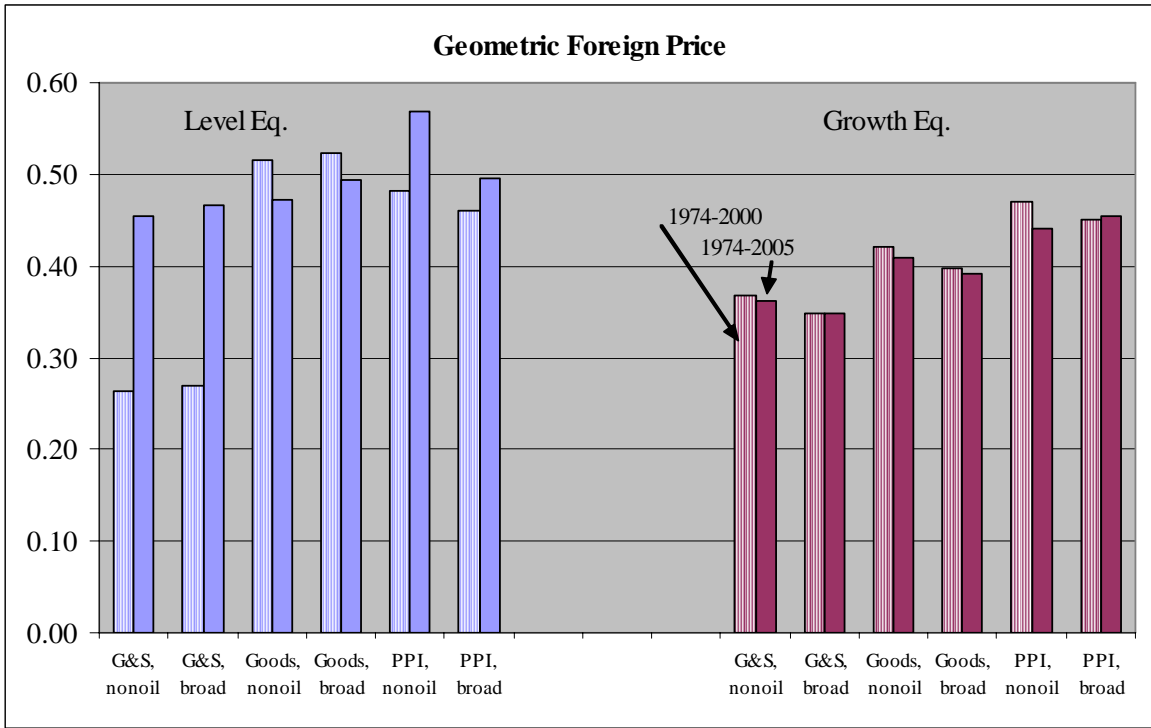


Figure 5: Long-run Estimates of Exchange-rate Pass-through ($\hat{\beta}$): Alternative Sample Periods

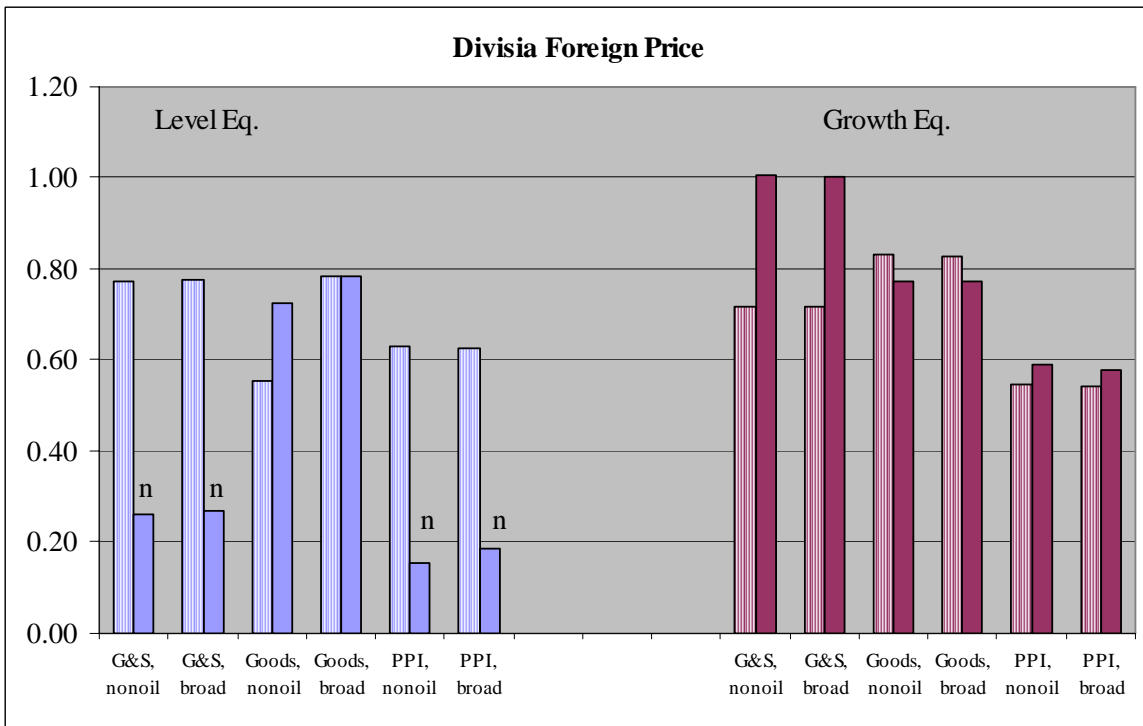
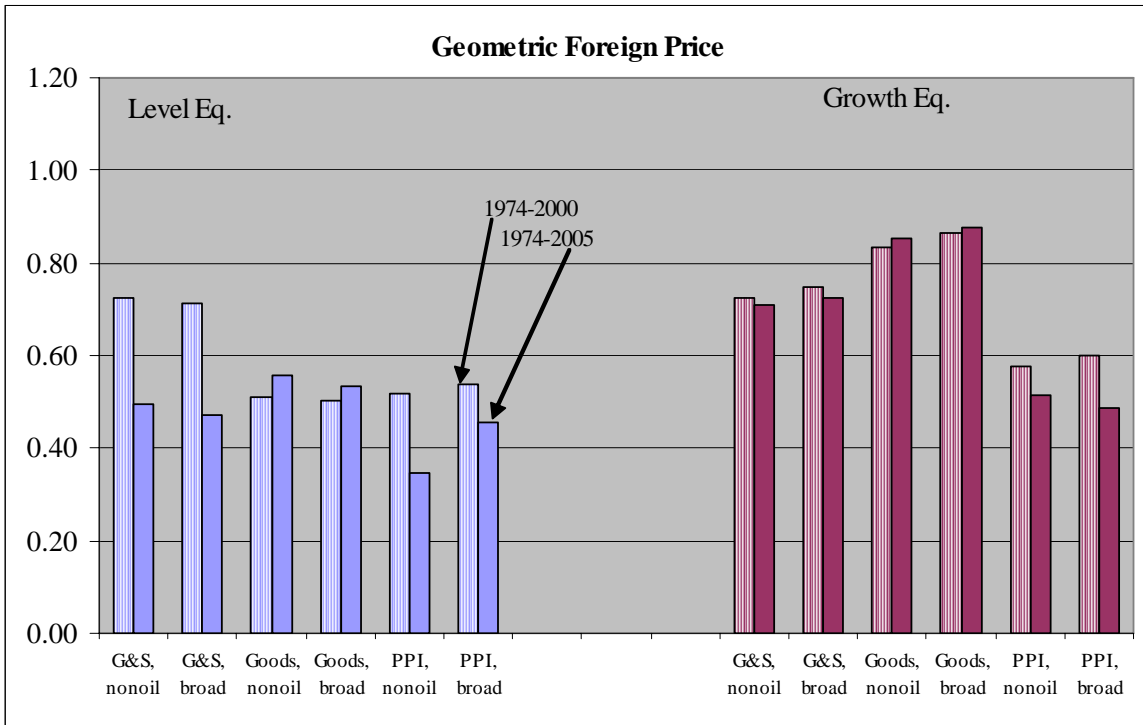


Figure 6: Long-run Estimates of Pricing-to-Market ($\hat{\gamma}$): Alternative Sample Periods (n denotes not stastically significant).

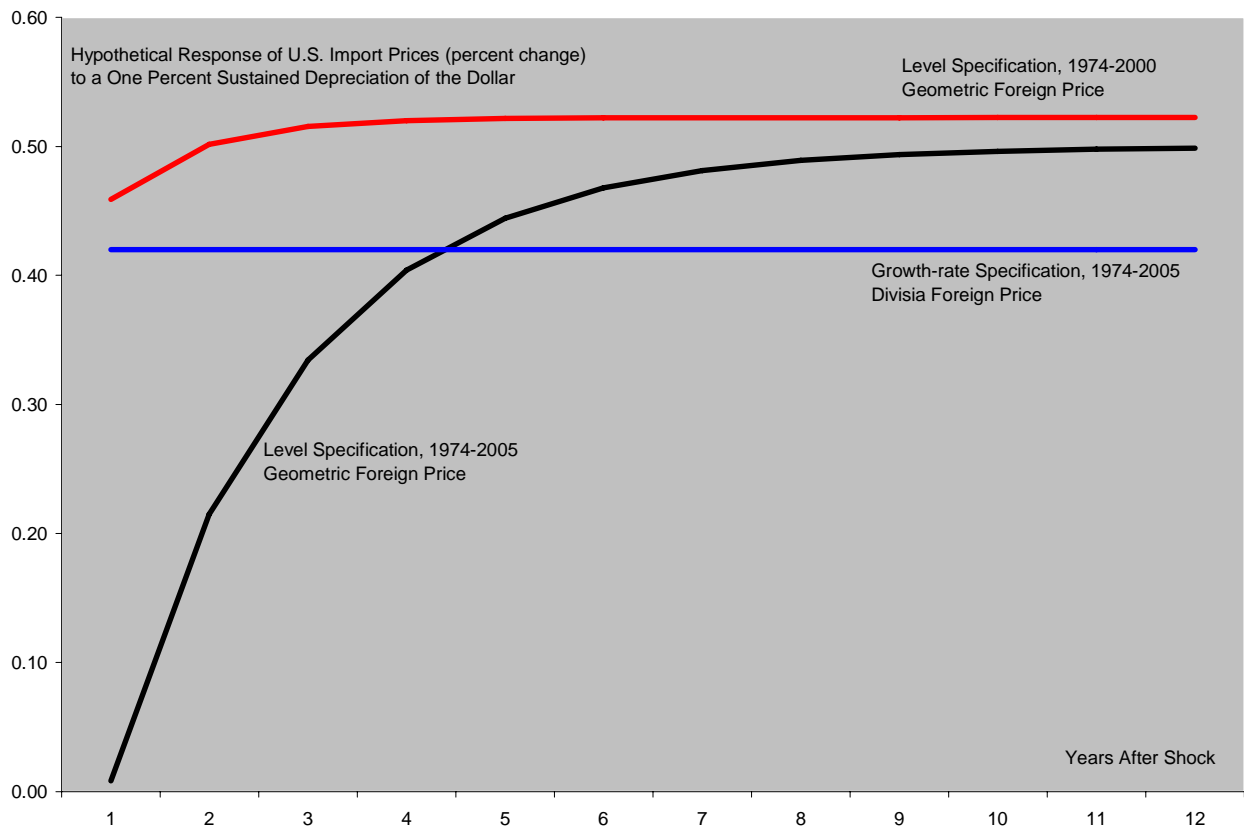


Figure 7: Short- and Long-run Exchange-rate Pass-through: Alternative Formulations

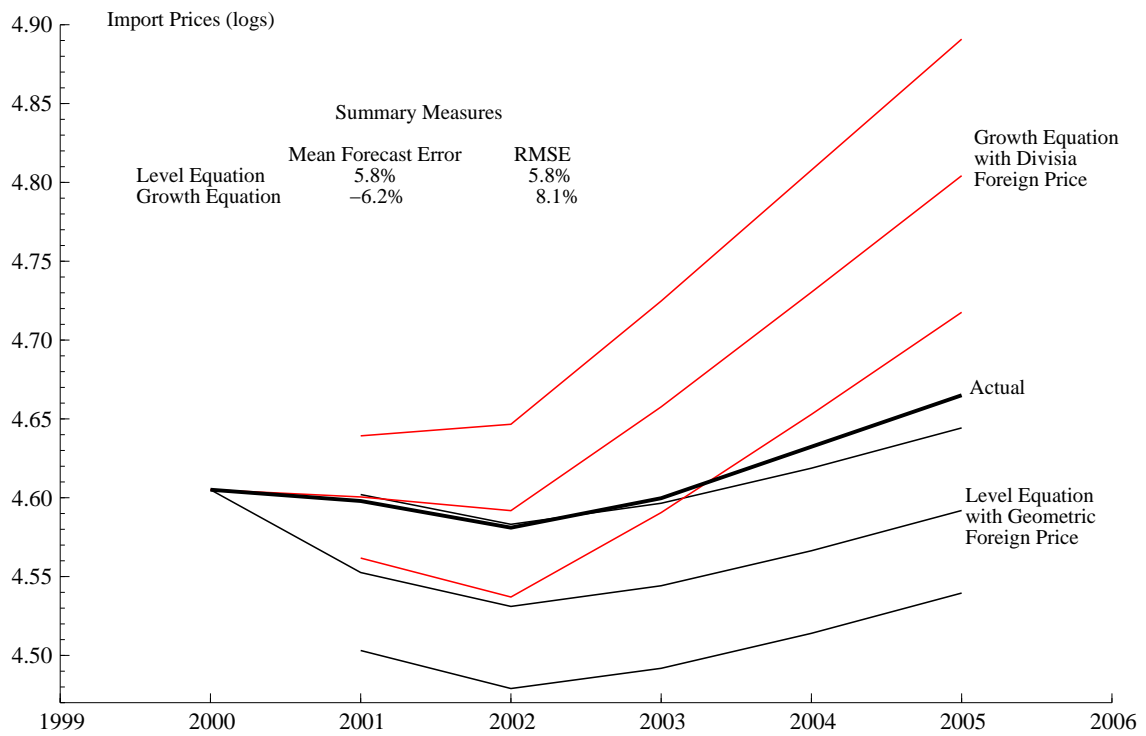


Figure 8: 95% Confidence Bands for Ex-post Dynamic Extrapolations of Import Prices

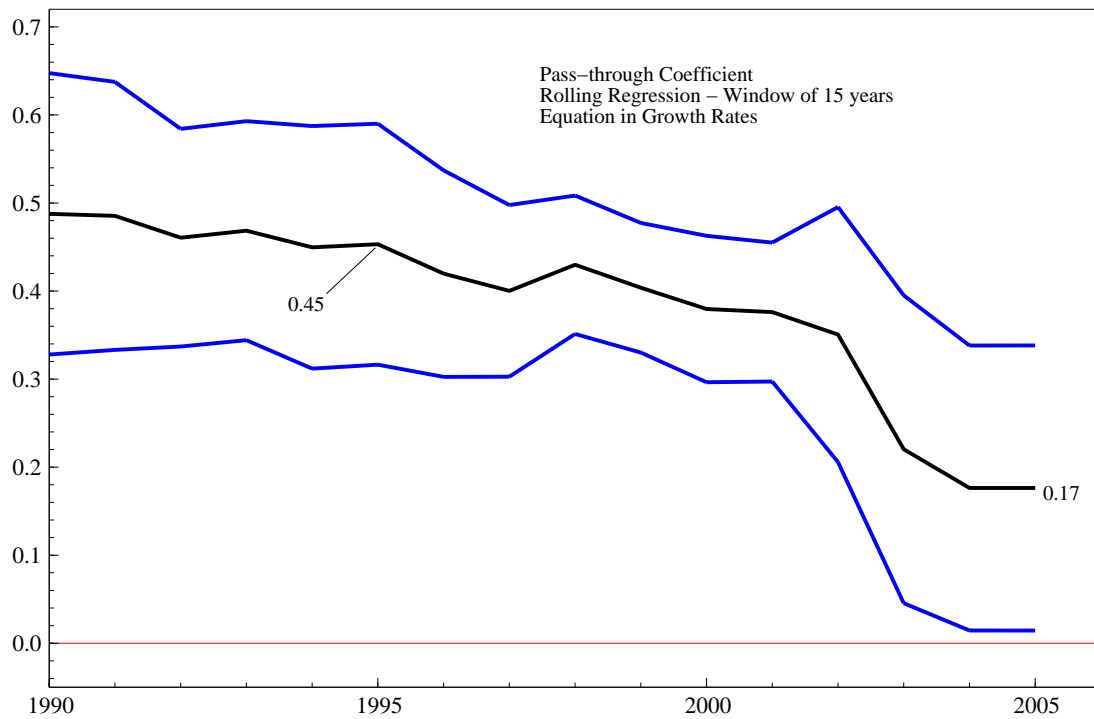


Figure 9: Estimated Short-run Exchange-Rate Pass-through: Growth-Rate Model with Rolling Regressions

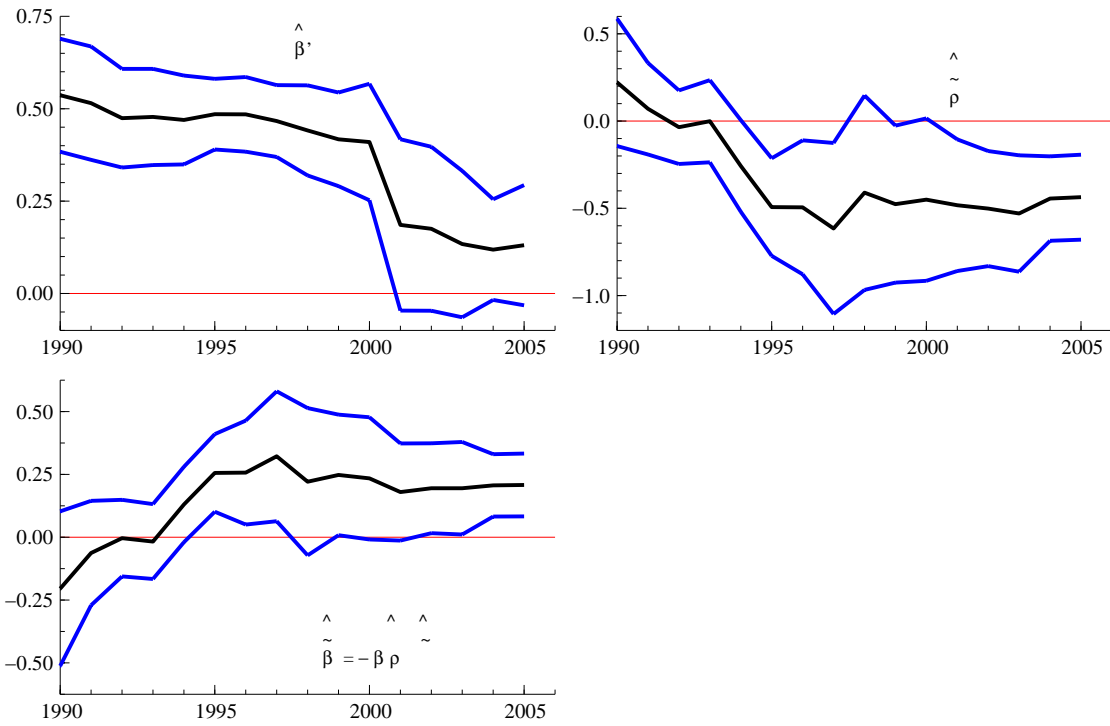


Figure 10: Rolling Regression Estimates of Error-Correction Model

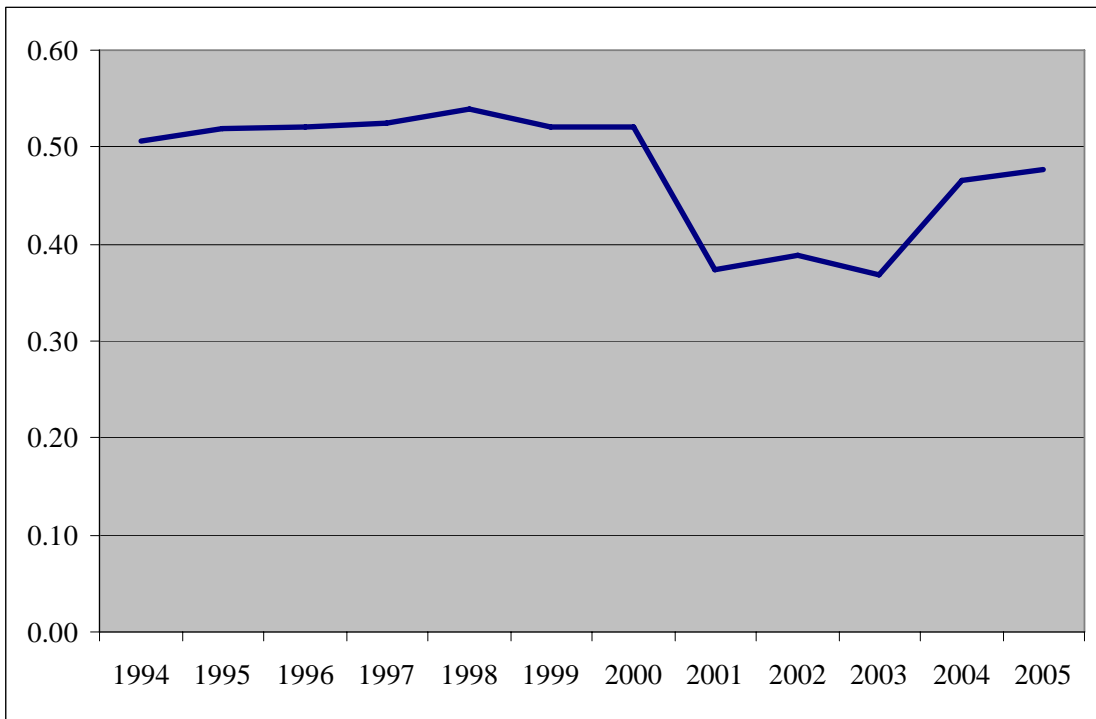


Figure 11: Long-run Estimate of Exchange-Rate Pass-through of Error-Correction Model: $-\frac{\partial \beta}{\partial \rho}$

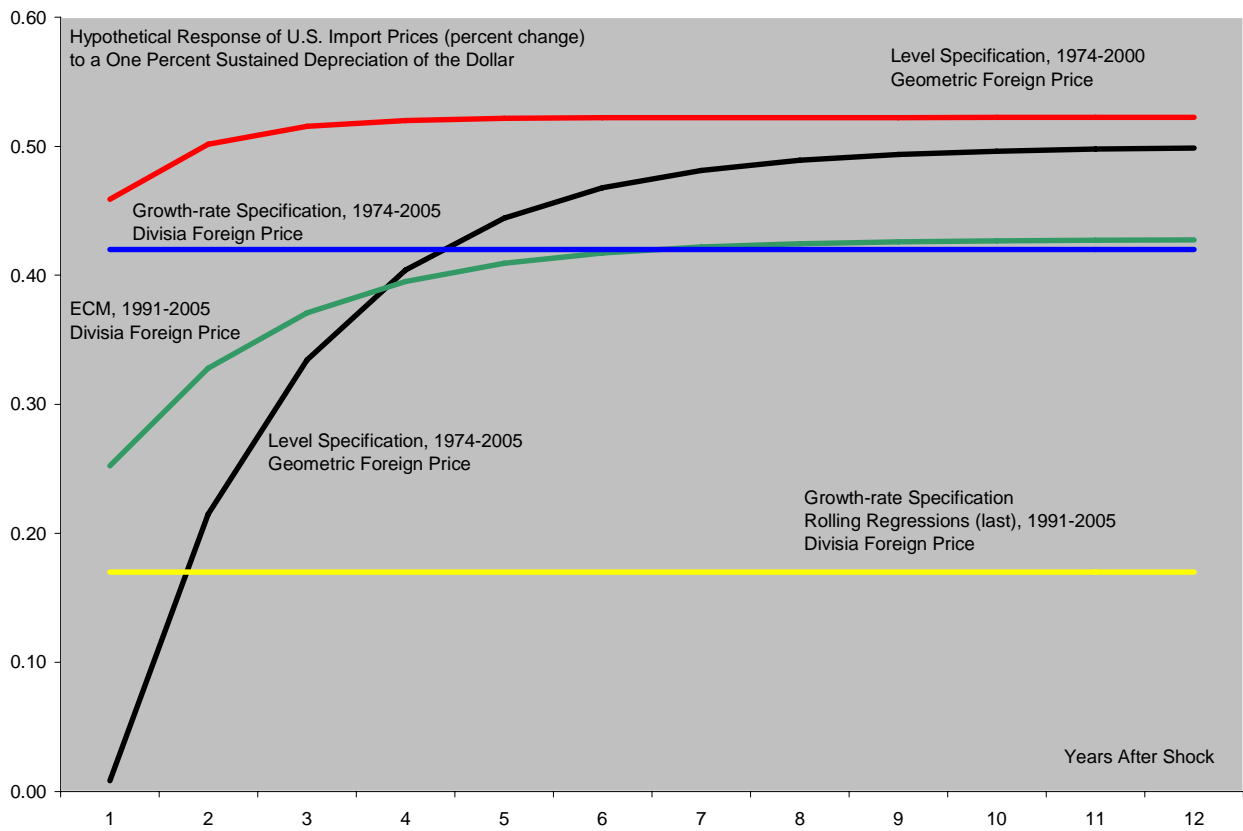


Figure 12: Short- and Long-run Exchange-rate Pass-through: Alternative Formulations and Sample-selection Methods