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## Measuring U.S. International Relative Prices: A WARP View of the World

Charles P. Thomas, Jaime Marquez, and Sean Fahle\*

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**Abstract:** In this paper we construct a new measure of U.S. prices relative to those of its trading partners and use it to reexamine the behavior of U.S. net exports. Our measure differs from existing measures of the dollar's real effective exchange rate (REER) in that it explicitly incorporates both the difference in price levels between the United States and developing economies and the growing importance of these developing economies in world trade. Unlike existing REERs, our measure shows that relative U.S. prices have increased significantly over the past 15 years. In terms of simple correlations, the relationship between our measure of relative prices and U.S. net exports is much more coherent than that between existing REERs and net exports. To explore this relationship further, we use our measure to construct an index of foreign prices relevant for U.S. export volumes and reexamine several export equations. We find that export equations with the new index dominate those with previous measures in terms of in-sample fit, out-of-sample fit, and parameter constancy. In addition, we find that with the new index of foreign prices the estimated elasticity of U.S. exports with respect to foreign income is a good bit higher than the unitary elasticity found in previous studies using other price measures. This has implications for U.S. current account adjustment.

**Keywords:** Automated Model Specification, China, Competitiveness, IMF, FRB, Geometric Aggregation, Penn World Tables, Real Effective Exchange Rates, Trade Elasticities

**JEL Classification:** C82, F41, C22

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# 1 Introduction

In this paper we assemble a new measure of international relative prices to gauge the average amount by which U.S. prices differ from foreign prices. Interest in developing such measures in international economics is not new.<sup>1</sup> What is new in this paper is the focus on the interactions between the dispersion of prices across countries and the increased trade with emerging economies. Recognition of these interactions yields a picture of U.S. international relative prices that is fundamentally different from the one given by existing measures of the real effective exchange rate. Indeed, unlike existing measures of relative prices, we find a significant increase in U.S. prices relative to its trading partners over the past 15 years. Further, most of this increase owes to greater trade with developing economies rather than increases in U.S. prices relative to individual countries.

Our measure differs from those currently available for two reasons. First, we measure bilateral relative price *levels*, as opposed to bilateral relative price *indexes*. Second, we use an aggregation method that retains the information embodied in those levels. In contrast, existing measures of relative prices are constructed by either chaining or averaging indexes – that is, they begin with price and exchange rate indexes constructed to have a value of 100 in a base year so that the value of the index in a given period indicates how much prices have changed since the base year. Thus, multi-country aggregates of these indexes measure the average change relative to the base year. Such methods are ideal if the purpose is to measure average changes in bilateral real exchange rates but not for measuring the level of U.S. prices relative to prices elsewhere.

Of course, others have recognized the importance of differentiating price indexes from price levels.<sup>2</sup> But the implications of combining that distinction with the increased role of developing countries in world trade has not received attention. In particular, the fact that prices in some developing economies are systematically below those in developed economies, combined with the fact that emerging economies' share of world trade has been increasing, has led to a decline in the average world price of traded goods even though prices in individual countries have not fallen. Aggregates based on price indexes cannot capture this interaction between price levels and trade shares. Our weighted average relative price (WARP) is designed specifically to capture this interaction and does so by using a geometric aggregate where the weights capture the change in the structure of U.S. external trade.

Section 2 reviews the evolution of several well known real effective exchange rate indexes. Al-

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<sup>1</sup>For a recent review, see Froot and Rogoff (1995). For early work on the importance of measuring relative prices, see Keynes (1925), Kravis and Gilbert (1954), and Kravis and Lipsey (1971). Other relevant papers include Lipsey, Molinari, and Kravis (1990), Hooper and Richardson (1991), and Turner and Van't dac (1993).

<sup>2</sup>For example, Turner and Van't dac (1993) examine this distinction using cross-sectional data.

though these indexes differ in source data and aggregation scheme, they generally paint a similar picture: U.S. prices relative to foreign prices have risen and fallen since 1975 but, on balance, they show no trend. Section 3 presents the WARP, discusses a few of its properties, and compares it to other measures. According to WARP, U.S. prices have risen significantly relative to its trading partners' prices since 1975 with most of the increase occurring since 1990.

This upward trend in U.S. international relative prices constitutes the main result of this paper. Section 4 examines several factors responsible for this upward trend: choice of price data, aggregation method, and currency basket. We find that the upward trend owes to the aggregation of relative price levels as such and to the shift in U.S. trade patterns away from the relatively high-price industrial countries toward the lower-price developing economies. Section 5 examines the sensitivity of this upward trend to both parametric structures and measurement errors; we find that the upward trend of U.S. international relative prices is robust.

Section 6 addresses whether WARP can be thought of as a measure of competitiveness. A point that comes clearly from the analysis is that any reasonable measure of competitiveness will necessarily incorporate the prices of non-traded goods and services as well as the prices of traded goods and services. Indeed, with analytical examples we show why the suitability of a measure of competitiveness to a particular application is largely an empirical question. With this in mind, Section 6 also looks at the relationship between WARP and the U.S. trade balance. We find that in terms of simple correlations, the relationship between relative prices and the U.S. trade balance (as a share of GDP) is much tighter when one uses WARP than when using conventional measures of real effective exchange rates. To explore why this might be the case, we examine several econometric specifications for the volume of U.S. exports. The focus is on assessing the implications for parameter estimates of using WARP-based and other measures of foreign prices to construct a relative price of exports. Our goal is not to offer detailed specifications for exports but, rather, to see if the WARP passes the "proof of concept" test. The evidence suggests that it does.

## 2 Existing Measures of Relative Prices

Existing measures of the dollar's real effective exchange rate (REER) are designed to reflect how much, on average, U.S. prices have *changed* relative to the prices of its trading partners.<sup>3</sup> The top panel of figure 1 shows the measures constructed by the Federal Reserve, the OECD, and the IMF,

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<sup>3</sup>The theoretical underpinnings of the REERs date back to work by Armington who, as McGuirk (1986, p. 3) points out, showed that an ideal weighting system is one in which an equiproportionate change in the product prices of all countries would leave the demand for any one country's product unchanged.

all of which are based on relative CPIs.<sup>4</sup> Though they differ from one another in many important methodological respects, they all show two common features. First, over the past thirty years, U.S. relative prices have changed little on average, a property that is at odds with the growing U.S. current-account deficit. Second, over shorter periods, U.S. international relative prices deviate substantially from their long-term mean and indeed these prices reached a historical peak in 1985.<sup>5</sup>

These three measures are constructed by aggregating bilateral real exchange rate indexes. That is, they begin with bilateral nominal exchange rate indexes and adjust them by relative movements in U.S. and foreign consumer price indexes. These bilateral real exchange rates (indexed to 1973=100) are shown in the middle panels of figure 1. The left panel plots the indexes vis-a-vis selected industrial countries; the right panel plots the indexes vis-a-vis selected emerging economies.<sup>6</sup> There is clearly a good deal of dispersion among these bilateral indexes, indicating that the CPI-adjusted value of the dollar has risen relative to some countries' currencies and fallen relative to others. On a bilateral basis, these real exchange rates can be interpreted as changes in relative prices.

Given the dispersion of bilateral real exchange rates across countries, it is hard to tell if there is a general pattern to the movements. This is the point of a REER: to distill these various movements into a single measure. To do so requires a weighting scheme. The aggregates shown in the top panels use weights based on trade shares. The weights used by the Federal Reserve Board in its Broad Real Index are representative; a selection of these is given in the bottom panels of figure 1.<sup>7</sup> We note the increasing weight given to developing economies, especially China and Mexico, since 1990.

How has the increased weight of the developing economies affected the REERs? If one looks at both the increase (depreciation) in China's bilateral real exchange rate since 1973 (middle right panel) and the increase in China's weight in U.S. trade since 1990 (bottom left) one might conclude that China's real exchange rate has had a significant impact on the dollar's REER. However, in fact, China's real exchange rate has had a relatively small effect on the dollar's REER. The mechanics for this result vary with the particular REER used but, in general, the reason is that most of the

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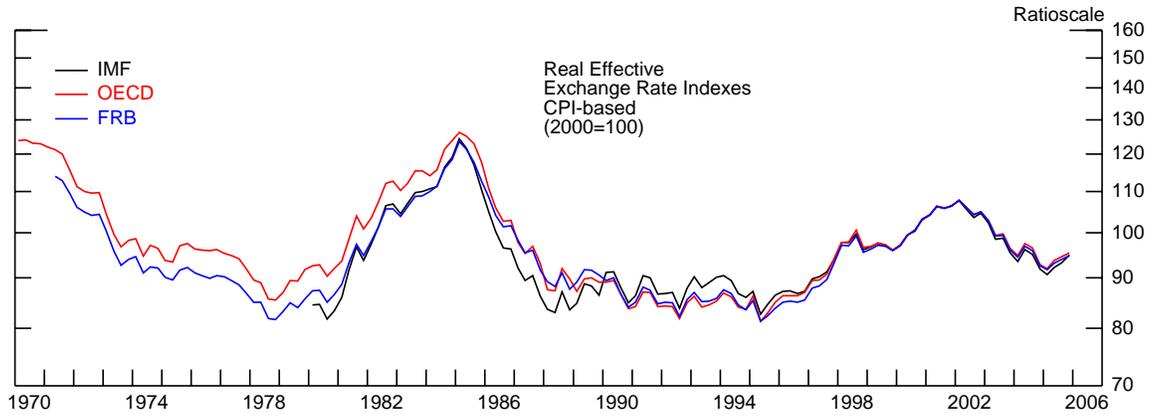
<sup>4</sup>Both the OECD and the IMF also report real effective exchange rates that are based on unit-labor costs; these measures show pronounced secular declines.

<sup>5</sup>Chinn (2005) reviews these measures. For the IMF, see Bayoumi et al. (2005), Zanello and Desruelle (1997), Turner and Golub (1997), Maciejewski (1983); for the Federal Reserve, see Hooper and Morton (1978), Pauls (1987), Leahy (1998), Loretan (2005); for the OECD, see Durand, Simon, and Webb (1992), Durand, Madaschi, and Terribile (1998). The BIS also constructs a real effective exchange rate comparable to that of the OECD and the Federal Reserve but the series starts in 1994 and so it is not suitable for our analysis; see Klau and Fung (2006).

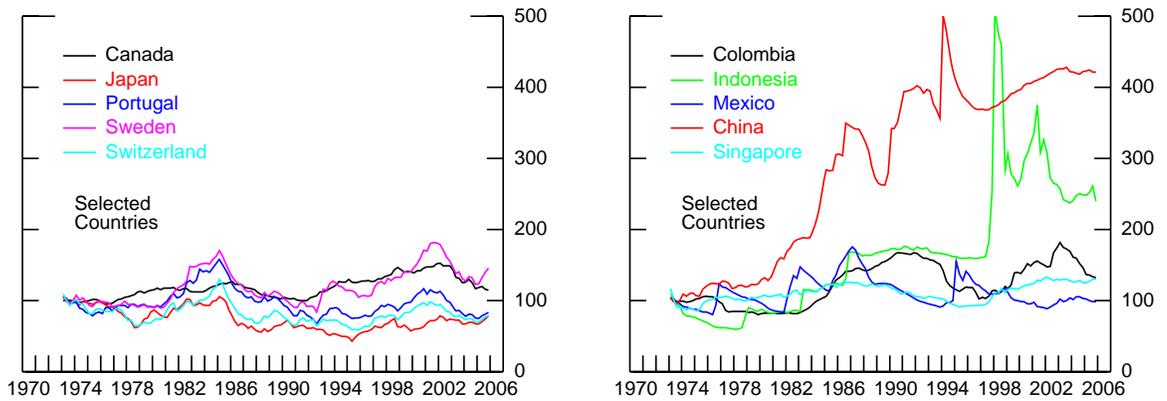
<sup>6</sup>The countries for each group were selected so as to encompass (form an envelope around) all the bilateral real exchange rates in our sample of 34 countries. The group of industrial countries corresponds to the Federal Reserve's classification of countries with major currencies; the group of emerging economies corresponds to the Federal Reserve's list of currencies of Other Important Trading Partners (OITP). These country groupings coincide, respectively, with the "High-price" and "Low-price" countries.

<sup>7</sup>The FRB's weight for a given country consists of trade shares for bilateral non-oil imports, bilateral exports, and a measure of the importance of the competition between that country and the United States in third-country markets. For further details on the construction of these weights, see Leahy (1998).

### Real Effective Exchange Rates



### Bilateral Real Exchange Rate Indexes - CPI-based (1973=100)



### Aggregation Weights (Federal Reserve Board's Measure)

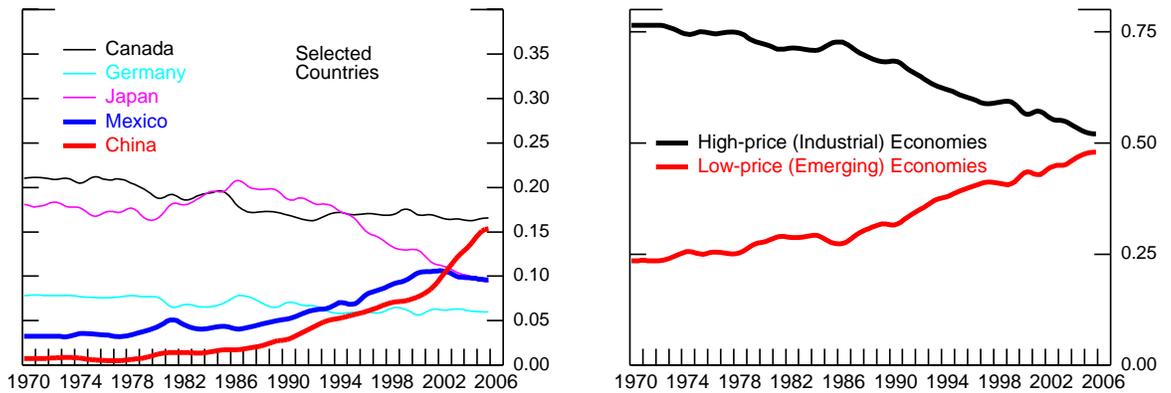


Figure 1: Real Exchange Rate Indexes: Effective and Bilateral – Selected Institutions

increase in the dollar real bilateral exchange rate vis-a-vis China occurred prior to 1990, a time when China's weight in U.S. trade was relatively small. REERs designed to show average changes do not get much of a boost from the Yuan's real depreciation prior to 1990 because China's weight in the index was small during that period. Conversely, despite the increase in the weight of China after 1990, there has not been much real depreciation of the Yuan during the period when the weight was large, so, again, the REERs do not get much of a boost. In general, what matters for existing measures of real effective exchange rates is whether the bilateral exchange rates are changing and, if they are not changing much, then increasing the weights on these countries does not cause the REER to change.

If the sole objective is to measure *changes* in the real effective exchange rates, then one can hardly improve upon existing measures. What we argue is that an exclusive focus on such changes carries a loss of information, that this loss is more than a theoretical possibility, and that the increased participation of low-cost producers in the world economy gives it economic significance.

### 3 The WARP

#### 3.1 Intuition

The basic idea behind our aggregate is simple. Suppose, for expository ease only, that we have the foreign-currency price of a basket of goods in a foreign country  $i$  (call it  $P_i$ ), and that we also have the dollar price of the same basket in the United States (call it  $P_{us}$ ). As shown in equation (1), by multiplying the ratio of these prices by the market exchange rate we can define a bilateral relative price  $q_i$  as

$$q_i = \frac{P_{us}}{P_i} \cdot E_{i/\$}. \quad (1)$$

This relative price is unitless and easy to interpret: A value of 2 means that the basket is twice as expensive in the United States as it is in country  $i$ .

To combine these bilateral relative prices into an aggregate measure for the United States, we use a weighted geometric mean where the weights vary over time and reflect each country's importance in U.S. trade. Specifically,

$$Q_t^g = (q_{1t})^{w_{1t}} \cdot (q_{2t})^{w_{2t}} \cdot \dots \cdot (q_{nt})^{w_{nt}}, \quad (2)$$

where  $w_{it}$  is the time-varying weight associated with the  $i$ th country.<sup>8</sup> Two features of  $Q^g$  are worth

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<sup>8</sup>Because  $Q^g$  is a limiting case of the more general CES function, section 5.1 below examines the sensitivity of our results to alternative parameterization.

noting. First, the level of the aggregate has meaning: a value of 1.5 means that U.S. prices are on average fifty percent above foreign prices and this value is not arbitrarily determined by the choice of base year. Second, the aggregate can change even if all bilateral relative prices are fixed.

An obvious alternative to  $Q^g$  is the commonly used chained aggregate, which is a weighted average of the growth rates of bilateral relative prices:

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

By convention,  $Q_{t=base}^c$  is set equal to 100 in a given base period and the level of the index for all other periods is defined recursively. Chained aggregation has two important features to recommend it. First, the index is independent of the levels of its constituent  $q$ 's, implying that we do not have to choose a meaningful base period for them. Second, changes in the aggregate index only reflect changes in the underlying relative prices. That is, if these rates do not change over a given period, then the aggregate index will not change, even if the weights do. Thus,  $Q_t^c$  may be ideal for measuring the average change in the dollar's bilateral relative prices.<sup>9</sup>

Given these aggregation formulas, how can, in the aggregate, U.S. prices rise relative to foreign prices? Holding all else equal, there are four channels:

1. The nominal dollar exchange rate can appreciate.
2. U.S. prices can rise.
3. Foreign prices can fall.
4. The weight of relatively low-price foreign economies can increase.

The first three channels operate through their impact on the bilateral relative prices – the  $q$ 's – and they are fully captured in both the geometric and chained aggregates. However, the chained index does not attempt to capture the fourth channel whereas the geometric aggregate does so explicitly. Specifically, logarithmic differentiation of equations (2) and (3) with respect to time yields

$$\begin{aligned} d \ln Q_t^g &= \sum_i w_{it} \cdot d \ln (q_{it}) + \sum_i dw_{it} \cdot \ln (q_{it}) \\ d \ln Q_t^c &= \sum_i w_{it} \cdot d \ln (q_{it}), \end{aligned}$$

which implies that

$$d \ln Q_t^g - d \ln Q_t^c = \sum_i dw_{it} \cdot \ln (q_{it}).$$

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<sup>9</sup>Appendix A.2 documents the properties of  $Q^g$  and  $Q^c$  using numerical examples.

Thus the difference in growth rates between the geometric and the chained aggregate is  $\sum_i d\omega_{it} \cdot \ln(q_{it})$ . This term captures the interaction between each period’s distribution of the level of bilateral relative prices and the evolution of the weights; if the weights are constant, then the two growth rates are identical.

## 3.2 Implementation

The previous discussion assumed, for expository convenience, the availability of data for the price levels of the foreign and domestic baskets. Thus the first step in implementing our measure is to obtain the bilateral relative prices—the  $q$ ’s. Data for bilateral relative prices are particularly difficult to obtain because they require comparability of products across countries.<sup>10</sup> To this end we use the Penn World Tables, which offer data on purchasing power parties.<sup>11</sup>

Greatly simplified, Penn collects data on spending and prices for products that are comparable across countries to estimate bilateral purchasing power parities. To avoid the calculations being sensitive to the choice of base country, Penn introduces the concept of “international dollars.” This strategy generates a system of simultaneous equations—the Geary-Khamis system—in which the PPP estimates depend on the international dollar and vice versa.<sup>12</sup> Specifically, given the international-dollar price of the  $j$ th product,  $\pi_j$ , the purchasing power parity for the  $i$ th country is

$$PPP_{\frac{i}{s}} = \frac{\sum_{j=1}^m P_j^i \cdot Y_j^i}{\sum_{j=1}^m \pi_j \cdot Y_j^i}, \quad i = 1, \dots, n, \quad (4)$$

where  $P_j^i$  is the price of the  $j$ th product in the  $i$ th country,  $Y_j^i$  is the amount produced of the  $j$ th product in the  $i$ th country, and the  $j$  index runs over the list of goods and services included in GDP.<sup>13</sup> The numerator equals the nominal GDP of the  $i$ th country expressed in local-currency terms whereas the denominator is the value of  $i$ th country’s GDP expressed in international dollars.

<sup>10</sup>See Vachrís and Thomas (1999) for the importance of comparability.

<sup>11</sup>Penn’s purchasing power parties have been used extensively in empirical analyses for the last three decades. See Heston, Summers and Aten (2006). For an introduction to the Penn World Tables, see Summers and Heston (1991) and Gulde and Schulze-Ghattas (1993). For the associated details, see Kravis, Heston, and Summers (1978, 1982). Other institutions also report purchasing power parities, but we use those from Penn because they include a relatively long time series. For recent efforts to improve the Penn World Tables, see Feenstra et al. (2005).

<sup>12</sup>Our presentation follows closely that of Gulde and Schulze-Ghattas (1993); see Kravis, Heston, Summers (1978, 1982) for additional details.

<sup>13</sup>Note that the weights (the  $Y$ s) are the same for the international dollar and for the prices of the  $j$ th country. We want to emphasize that these  $Y$ s are not physical measures of output but notional quantities; see Kravis, Heston, and Summers (1978, 1982) for details of this concept.

Given  $PPP_{\$/}$ , the international dollar price for the  $j$ th product is computed as

$$\pi_j = \sum_{i=1}^n \frac{P_j^i}{PPP_{\$/}^i} \cdot \left( \frac{Y_j^i}{\sum_{i=1}^n Y_j^i} \right), \quad j = 1, \dots, m, \quad (5)$$

where the first term is the price of the  $j$ th product in the  $i$ th country expressed in international dollars and the second term is the  $i$ th's country share in world output of the  $j$ th product.

The system given by (4) and (5) consists of  $m + n$  equations, of which only  $m + n - 1$  are linearly independent. To address the over-determined character of the system, Penn uses the United States as the numeraire country meaning that the international dollar has the same purchasing power over total U.S. GDP as the U.S. dollar. Thus the average U.S. price relative to the average price of the  $i$ th country can be estimated as the market exchange rate divided by Penn's PPP:

$$q_i = \frac{E_{\$/}^i}{PPP_{\$/}^i}. \quad (6)$$

There are several drawbacks to the Penn data for studying the open-economy implications of movements of U.S. international relative prices. First, the data are released with long delays: the most recent release (release 6.2 in 2006) has data ending in 2004.<sup>14</sup> Second, the data are annual. To address these two limitations, the paper develops a method to extend Penn's annual parities and to estimate the associated quarterly observations.<sup>15</sup> Finally, the data are subject to errors and section 5 examines the implications of these errors for our measure of U.S. international relative prices.

With these considerations in mind, the top panels of figure 2 show the evolution of the levels of bilateral relative prices for selected countries.<sup>16</sup> Among the industrial countries (left panel), U.S. prices are highest relative to Portugal and lowest relative to Switzerland with most measures near or a little below one. As shown to the right, among emerging economies, there is a good deal more dispersion with relative prices ranging between 1.5 and 6.

For aggregation we use the same trade weights as those in the Federal Reserve's Broad Real Dollar index (shown in lower left panel). Note that between 1980 and 1990 the total weight of emerging economies held steady near 25 percent, but since 1990 it has doubled to near 50 percent, reflecting

<sup>14</sup>An earlier version of this paper used data from the 6.1 release and the results are quite comparable to the ones reported here. Nevertheless, recent data from the World Bank's International Comparison Program (December 2007) point to data revisions for recent years. Section 5.2 below examines the implications of measurement errors. Further, in personal communication, Alan Heston informed us that there will be a new release of the PWT. Once the new data are available, we will update our calculations.

<sup>15</sup>Appendix A.1 documents these methods; for estimating quarterly parities we impose the constraint that the average of quarterly parities for a given year must be equal to the annual Penn parity for that year.

<sup>16</sup>The countries for each group were selected so as to encompass all the bilateral relative prices in our sample.

### Bilateral Relative Price Levels

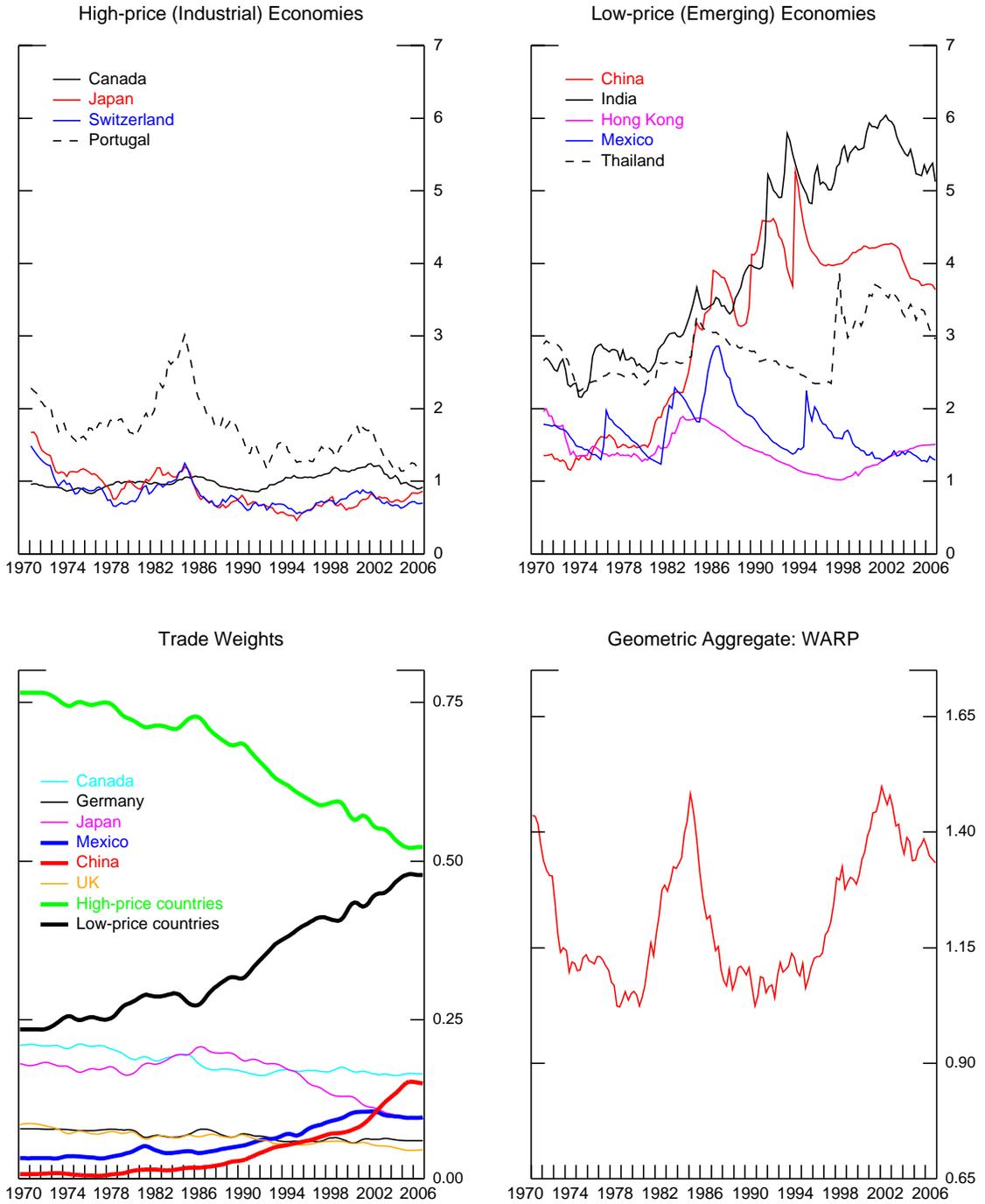


Figure 2: U.S. International Relative Prices: Bilateral and Aggregate

rising weights for China and Mexico. The weight for industrial countries has declined, with Japan's weight declining the most. The weighted average of the 34 bilateral relative prices is constructed using equation (2) and shown in the lower right panel. The aggregate of U.S. international relative prices shows an upward trend since the end of the Bretton-Woods period. Indeed, by this measure, U.S. prices are roughly 40 percent above those of its trading partners.

Figure 3 compares the evolution of this measure to the real effective exchange rates from the Federal Reserve and the IMF, rescaled by their own 1971-1991 sample means. The three measures move in near lockstep between 1971 and 1986. As such, neither the choice of aggregation method nor the measure of bilateral price has a noticeable effect on the aggregate measure of U.S. international relative prices through 1986. Since then, however, the aggregates tend to diverge. Specifically, the WARP shows a sustained increase and by 2002 it reaches the same value it had in 1985. In contrast, the other measures remain well below their 1985 peaks. This more recent divergence of U.S. international relative prices might be of interest in assessing the likelihood of a dollar depreciation large enough to address the U.S. external imbalance. Specifically, if one were to apply the 1985-1987 dollar depreciation to the 2006 values of the aggregates based on bilateral price indexes, then these aggregates would fall to levels not recorded in history. In contrast, applying the same depreciation to WARP would bring it to its 1986 value and, by this historical standard, such a depreciation would be consistent with previous experience.

## **4 Explaining the Rise in U.S. International Relative Prices**

We now look at why our WARP has risen much more than the other measures since the late 1980s. To ease the exposition, we abstract from differences involving country coverage and weighting scheme to focus on the measurement of bilateral prices and aggregation methods.

### **4.1 Aggregation Methods and Price Measures**

WARP differs from existing measures in both the choice of aggregation formula and the measure of bilateral prices, raising the question of which of these two factors explains the different trends in the aggregates. To address this question, we construct similar aggregates to those reported by other institutions where their bilateral relative prices are replaced with ours. This strategy ensures that any difference can be interpreted as due to the choice of aggregation method.

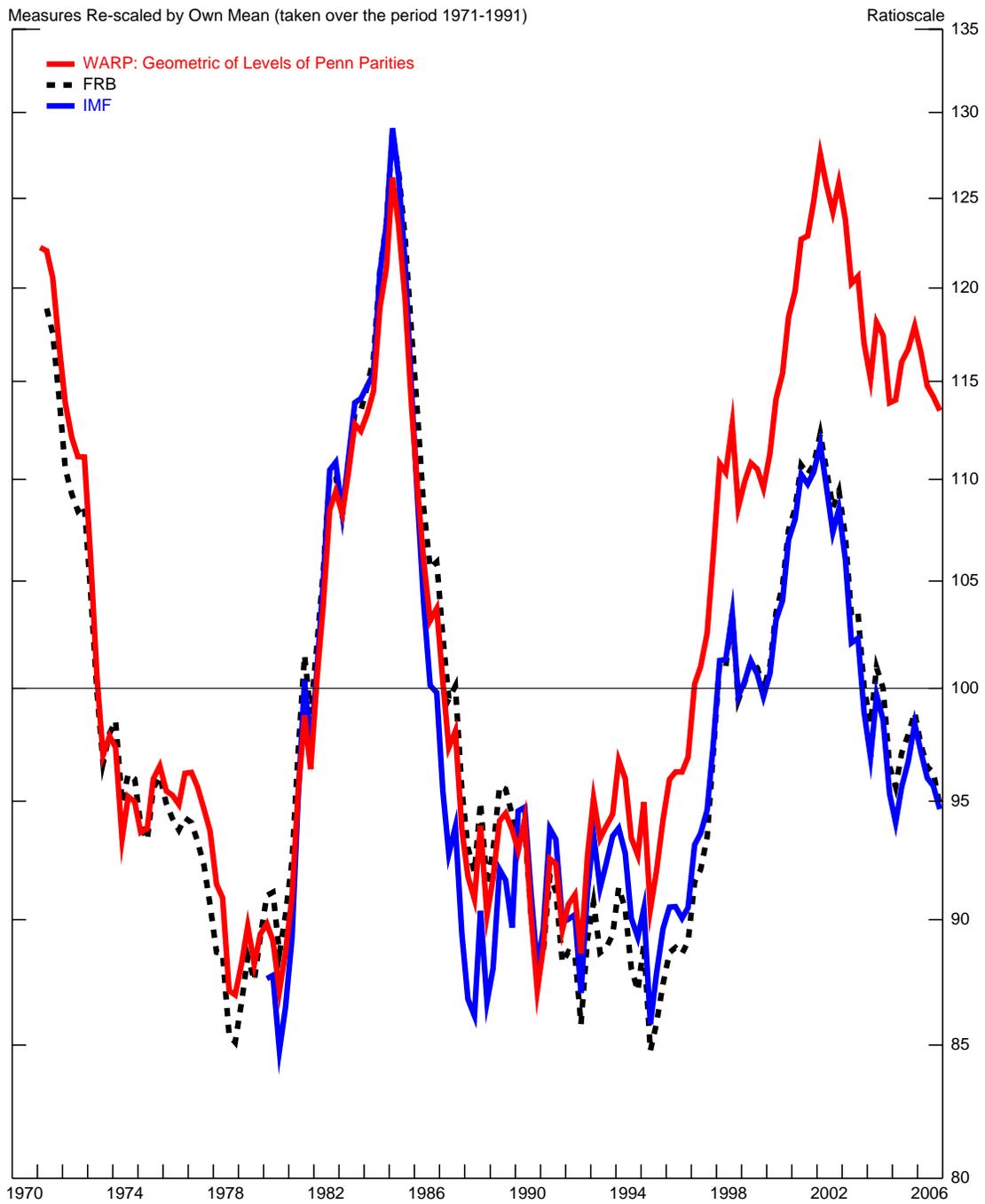


Figure 3: Alternative Measures of International Relative Prices

The Federal Reserve reports chained aggregates of bilateral CPI-adjusted exchange-rate indexes:

$$\frac{Q_t^c}{Q_{t-1}^c} = \prod_{i=1}^N \left( \frac{r_{\frac{i}{s},t}}{r_{\frac{i}{s},t-1}} \right)^{w_{it}}, \quad (7)$$

where

$$r_{it} = \left( \frac{CPI_{us,t}}{CPI_{i,t}} \right) \cdot \left( \frac{E_{\frac{i}{s},t}}{E_{\frac{i}{s},t_0}} \right),$$

$t_0$  represents the base period,  $CPI_{us}$  is the U.S. consumer price index, and  $CPI_i$  is the consumer price index for the  $i$ th country. The IMF reports a fixed-weight geometric aggregate of bilateral CPI-adjusted exchange rates:

$$Q_t^{gI} = \prod_{i=1}^N \left( r_{\frac{i}{s},t} \right)^{w_i}. \quad (8)$$

There are several differences between  $q_{it}$  and  $r_{it}$  that are potentially relevant for explaining differences between WARP and existing measures of real effective exchange rates. First,  $q_{it}$  measures the level of bilateral relative prices whereas  $r_{it}$  measures the percent change in bilateral relative prices. Second, the basket used for  $q_{it}$  refers to GDP items and thus includes consumption, investment, government purchases, and exports. The basket used for  $r_{it}$  is limited to consumption items both from domestic and foreign sources. Finally, the baskets embodied in  $q_{it}$  are the same for U.S. and foreign prices whereas the baskets embodied in  $r_{it}$  are not the same for U.S. and foreign prices.

To examine whether the upward trend in our WARP is due to differences between  $q_{it}$  and  $r_{it}$ , we construct the same aggregates reported by other institutions while using  $q_{it}$  instead of  $r_{it}$ . Following the FRB's methodology, we construct a chained aggregate of bilateral relative prices,  $Q_t^c$ , substituting the Penn parities (the  $q_{i's}$ ) into equation (7). We also report the geometric average of *indexes* of bilateral relative prices, similar to the IMF's methodology:

$$Q_t^{gI} = \prod_{i=1}^N \left( \frac{q_{it}}{q_{it_0}} \right)^{\omega_i},$$

where, following the IMF, we set  $q_{it_0}$  as the mean of the values of  $q_{it}$  in 2000 and  $\omega_i$  as the sample mean from 1989 to 1991.

Figure 4 shows the evolution of the real effective exchange rates from the Federal Reserve and the IMF along with  $Q_t^g$ ,  $Q_t^c$ , and  $Q_t^{gI}$ ; for comparison purposes, we rescale these measures by their own 1971-1991 sample means. The results indicate that aggregates based on Penn's bilateral relative price *indexes*,  $Q_t^c$  and  $Q_t^{gI}$ , show a downward trend meaning that the upward trend in WARP is not due to differences between  $q_{it}$  and  $r_{it}$  but, rather, to the choice of aggregation method.

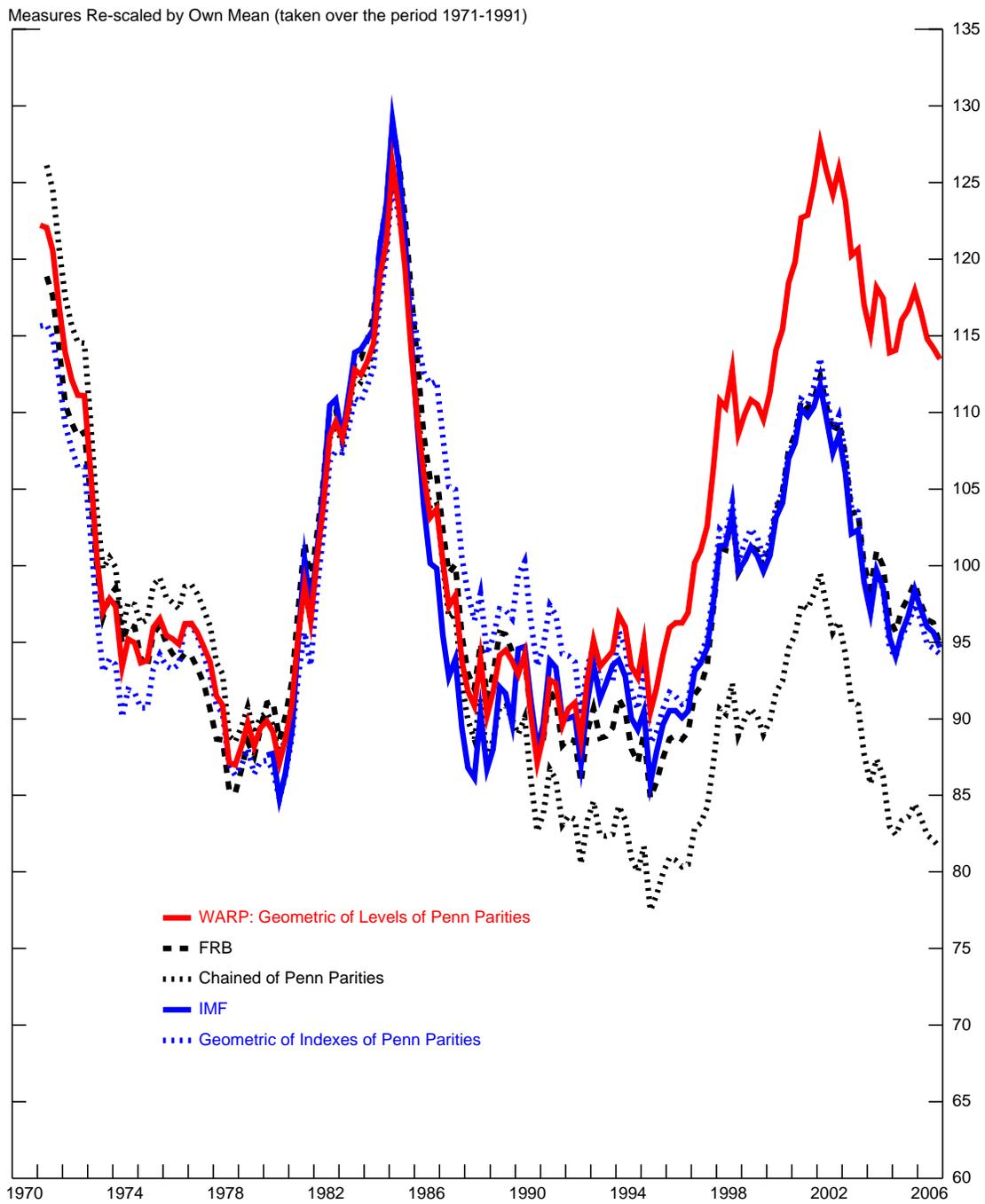


Figure 4: U.S. International Relative Price – Sensitivity to Price Data and Aggregation Method

## 4.2 Currency Baskets and Weighting Schemes

We now ask what factors in our aggregation method are responsible for the upward trend in  $Q^g$ . Is this trend due to the composition of the currency basket or to our weighting scheme? To address these questions, figure 5 reports separate geometric sub-aggregates for industrial countries and for emerging economies; the  $w'_{it}$ s for each group are renormalized to add up to one. The thick blue line is the aggregate of U.S. prices relative to other industrial countries alone. It has been trending down slowly, and it indicates that in 2006 U.S. prices were on average 10 percent below those in other industrial countries. The thick black line plots U.S. prices relative to the prices of emerging economies alone. It has been trending up sharply. These calculations suggest that a key factor accounting for the upward trend in our WARP, the thick red line, is the shift in U.S. trade patterns. Specifically, within the overall aggregate, trade has shifted away from the relatively high-price industrial countries toward the lower-price emerging economies, which tends to raise the overall measure of U.S. prices relative to our trading partners. Within the emerging economies sub-aggregate (the black line), trade has shifted toward the lowest-price economies, such as China; this shift tends to raise U.S. prices relative to the group.

To illustrate the importance of the increased weight of the low-price economies, we construct a counterfactual where we ask what would have happened if the weights after 1991 were fixed at their 1991 values. As shown by the dashed red line, in this world, our measure would have U.S. prices only about 10 percent above those of our trading partners—roughly unchanged since 1975 and near the 30-year average. Further, this fixed-weight aggregate has a downward trend with a historical peak in 1985, quite similar to the pattern of the standard aggregates shown earlier. The key question, however, is whose weights exert the strongest influence. To address that question, the figure reports the fixed-weight aggregate for the industrial countries. Appearing as the dashed blue line, this aggregate exhibits a downward trend with a historical peak in 1985, similar to the case of variable weights and to the associated aggregates reported by other institutions. This result implies that fixing industrial-country weights does not change the evolution of that sub-aggregate. In contrast, fixing the weights of each emerging economy induces a *downward* trend in the associated sub-aggregate (dashed black line), which leaves the historical peak back in 1985, unlike its variable-weight version which peaks in 2003. Therefore, the upward trend in our measure of U.S. international relative prices is due to the increased weight of the low-price economies in the U.S. basket.

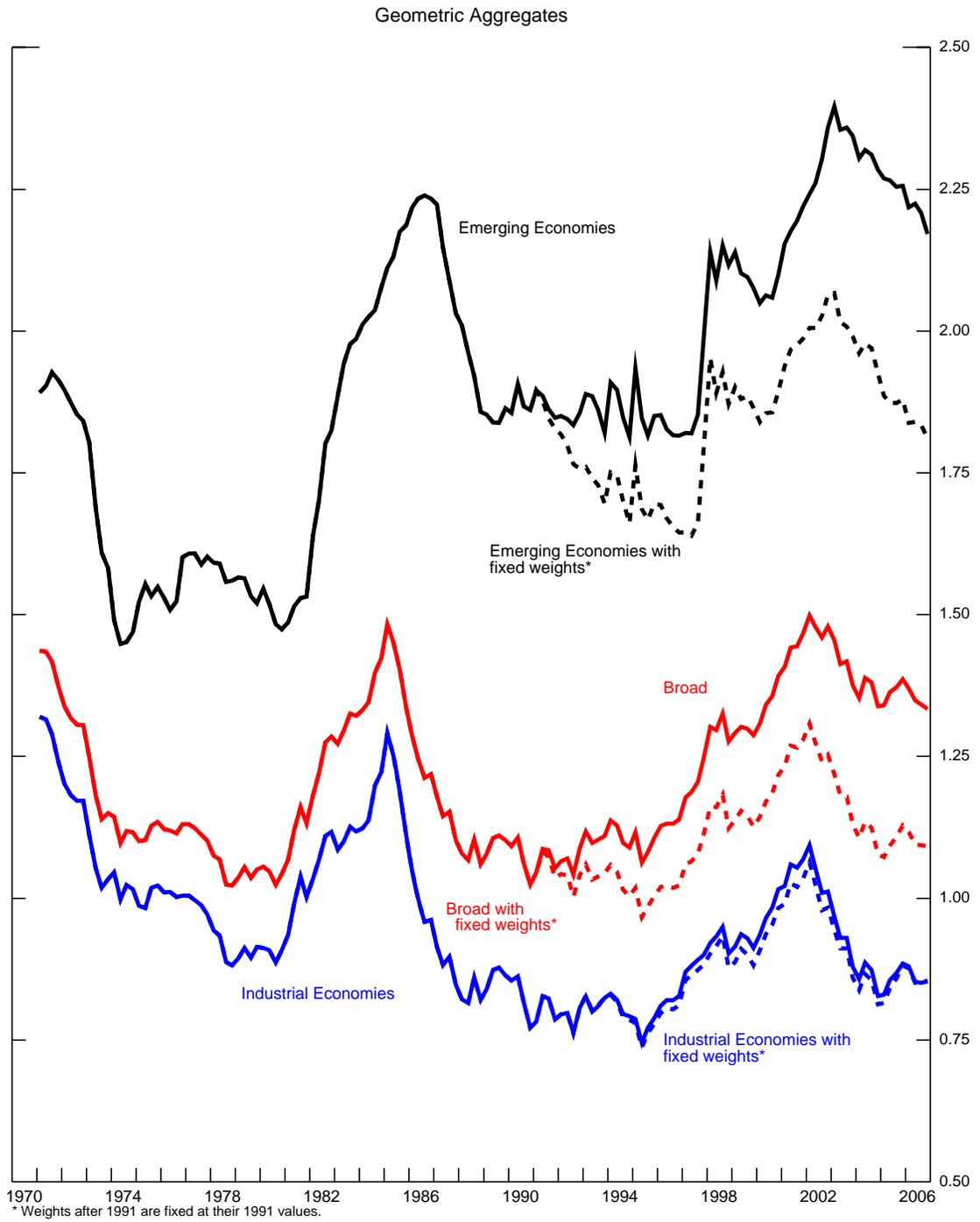


Figure 5: U.S. International Relative Price – Sensitivity to Currency Basket and Weighting Scheme

## 5 Sensitivity Analysis

### 5.1 Sensitivity to Parameterization

A well known result is that our  $Q^g$  is a particular case of the CES function

$$Q_t^{ces} = \left[ \sum_{i=1}^n w_{it} \cdot (q_{it})^{\sigma-1} \right]^{\frac{1}{\sigma-1}},$$

where  $\sigma$  is the elasticity of substitution among purchases of foreign products and  $Q^g = \lim_{\sigma \rightarrow 1} Q_t^{ces}$ . In the absence of econometric evidence supporting  $\sigma = 1$ , a relevant question to pose is how sensitive is the upward trend in U.S. international relative prices to alternative values of  $\sigma$ .<sup>17</sup>

Figure 6 shows the sensitivity of  $Q_t^{ces}$  to values of  $\sigma$  ranging from high substitutability ( $\sigma = 2.5$ ) to near complementarity ( $\sigma = 0.05$ ). The calculations reveal three findings. First, there is a direct association between the value of  $\sigma$  and the slope of the trend of U.S. relative prices. Second, the 2006 level of WARP is sensitive to extreme values of  $\sigma$  (2.5 and 0.05); using using less extreme values of  $\sigma$  (1.1 and 0.9) yields values of  $Q_t^{ces}$  quite close to the values taken by  $Q^g$ . Finally, if one interprets the large swings in U.S. bilateral trade shares as suggesting high substitutability among foreign products ( $\sigma > 2$ ), then the upward trend in  $Q^g$  understates the extent to which U.S. international relative prices have been increasing.

Overall, we interpret these results as suggesting that the upward trend in U.S. international relative prices associated with WARP is not due to our reliance on a unitary elasticity of substitution among foreign products. This conclusion, however, abstracts from the quality of the data of relative prices, an issue that we examine next.

### 5.2 Sensitivity to Measurement Errors

We now ask how sensitive is the upward trend of WARP to measurement errors. Other things equal, a lower estimate of  $PPP_{i/s}$  raises  $q_i$ , our measure of U.S. prices relative to country  $i$ . Thus a relevant question is whether we are over-estimating  $q_i$ , and hence WARP, because Penn is underestimating  $PPP_{i/s}$  for emerging economies. We consider three approaches to address this question: examining alternative measures of purchasing power parity for China; adjusting our measures of relative prices by imputing correction factors larger than those of Kravis and Lipsey (1990); and comparing the WARP to calculations based on the price data for the Big Mac.

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<sup>17</sup> Appendix A.3 shows the derivation of  $Q^{ces}$ .

Sensitivity to Degree of Differentiation

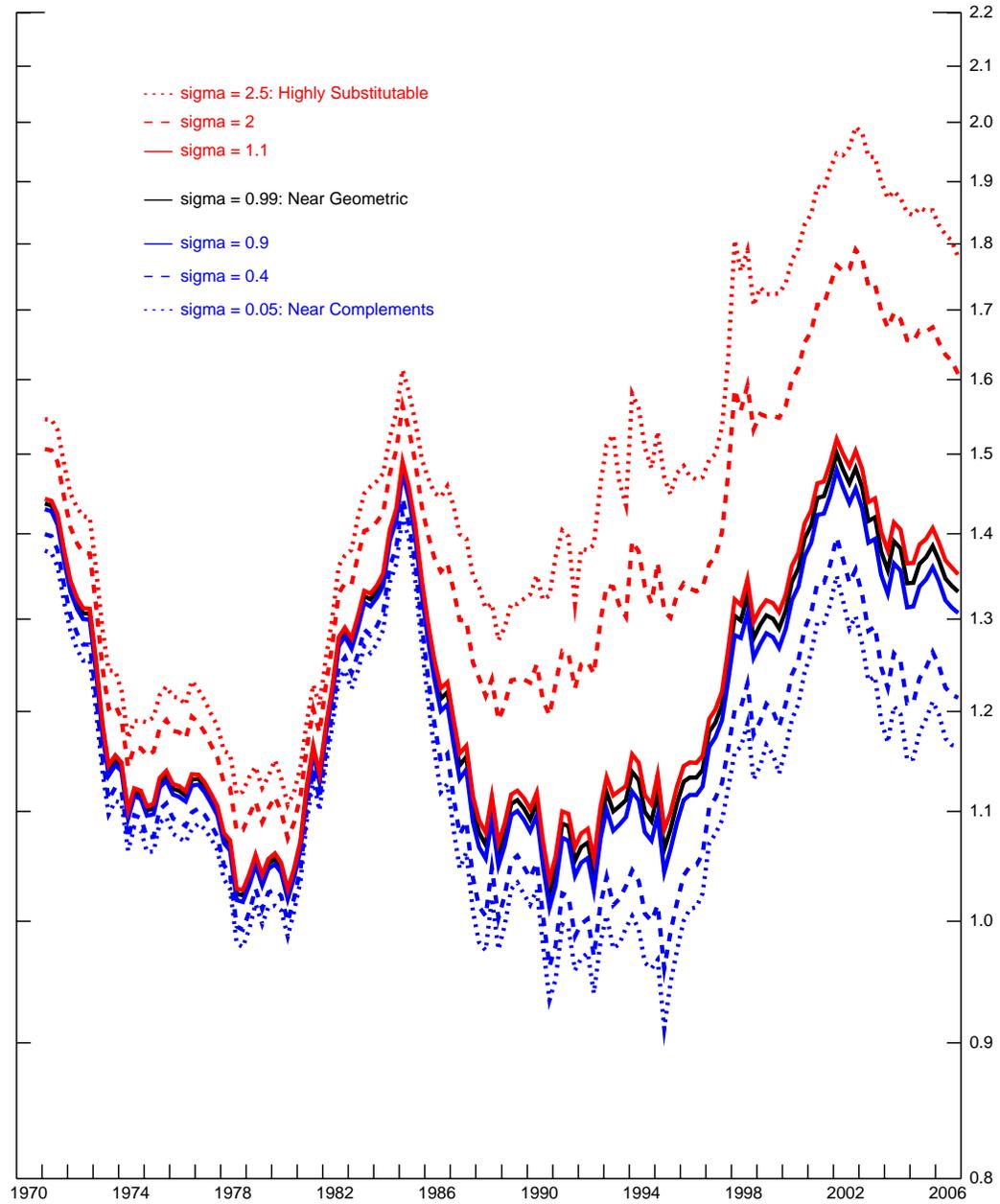


Figure 6: U.S. International Relative Prices – CES Aggregates with Alternative Values of  $\sigma$

### 5.2.1 Alternative Estimates of PPP for China

A focus on China's purchasing power parity can be motivated in two ways. First, the weight for China has experienced the largest increase and it now has the second largest value in our weighting scheme. Second, the price data for China are of questionable reliability. Figure 7 compares the estimates for China's  $PPP_{i/s}$  from Penn (solid dark-blue line) to the IMF's estimates from seven recent vintages.<sup>18</sup> Prior to 1994, Penn's estimate is never more than eight percent below the IMF estimates. For all the post-1994 period, Penn's estimate is at least as large as any of the estimates from the IMF. Thus there does not seem to be a systematic undervaluation of Penn's parities relative to those of the IMF. The one estimate we could find that is above that of Penn is that of the OECD.<sup>19</sup> For 2004, the OECD estimate for China is 2.3, compared to Penn's estimate of 2.1. Thus the OECD estimate is roughly 10 percent above the Penn estimate.

We do not interpret this scant evidence as suggesting that Penn's estimates are relatively error free but that, perhaps, comparing inaccurate measures across institutions is not informative. Thus we examine below the implications of imputing large measurement errors to the relative prices of emerging economies. Indeed, we find that even if the relative prices for all emerging economies were 20 percent below what Penn estimates, this is still not enough to overturn the basic upward trend of WARP.

### 5.2.2 Imputation of Correction Factors

We now impute measurement errors to the relative prices of emerging economies to examine the sensitivity of WARP to such mismeasurement. Specifically, we denote  $\tilde{q}_{it}$  as the error-free but unobserved bilateral relative price and postulate that  $\tilde{q}_{it} = (q_{it})^\theta$  where  $\theta$  is the imputed correction factor. We could impute the value of  $\theta$  using either the 13 percent estimated by Kravis and Lipsey (1990) or the 10 percent wedge implied by the OECD estimate for China. To encompass these sources and to allow for even larger errors, we apply  $\theta = 3/4$  to the relative prices of China, of Latin American countries, of all emerging economies excluding both China and Latin America countries, and of all emerging economies. Note that the magnitude of the error is directly related to the value of  $q$ . If  $q = 2$  and  $\theta = 3/4$ , then  $\tilde{q} = 1.68$  implying an error of 19 percent.

The top panel of figure 8 shows how U.S. prices relative to those of emerging economies respond to the imputed correction factor. We find that if  $\theta = 3/4$  is applied to all developing countries, then

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<sup>18</sup>The data come from the *World Economic Outlook*. The IMF's calculation starts with the PPP exchange rate in year 2000 from the World Bank; this value is then extended forwards and backwards by the growth in relative GDP deflators. See <http://www.imf.org/external/pubs/ft/weo/faq.htm#q21>

<sup>19</sup>See *OECD Economic Surveys China* Volume 2005/13 September 2005, page 9.

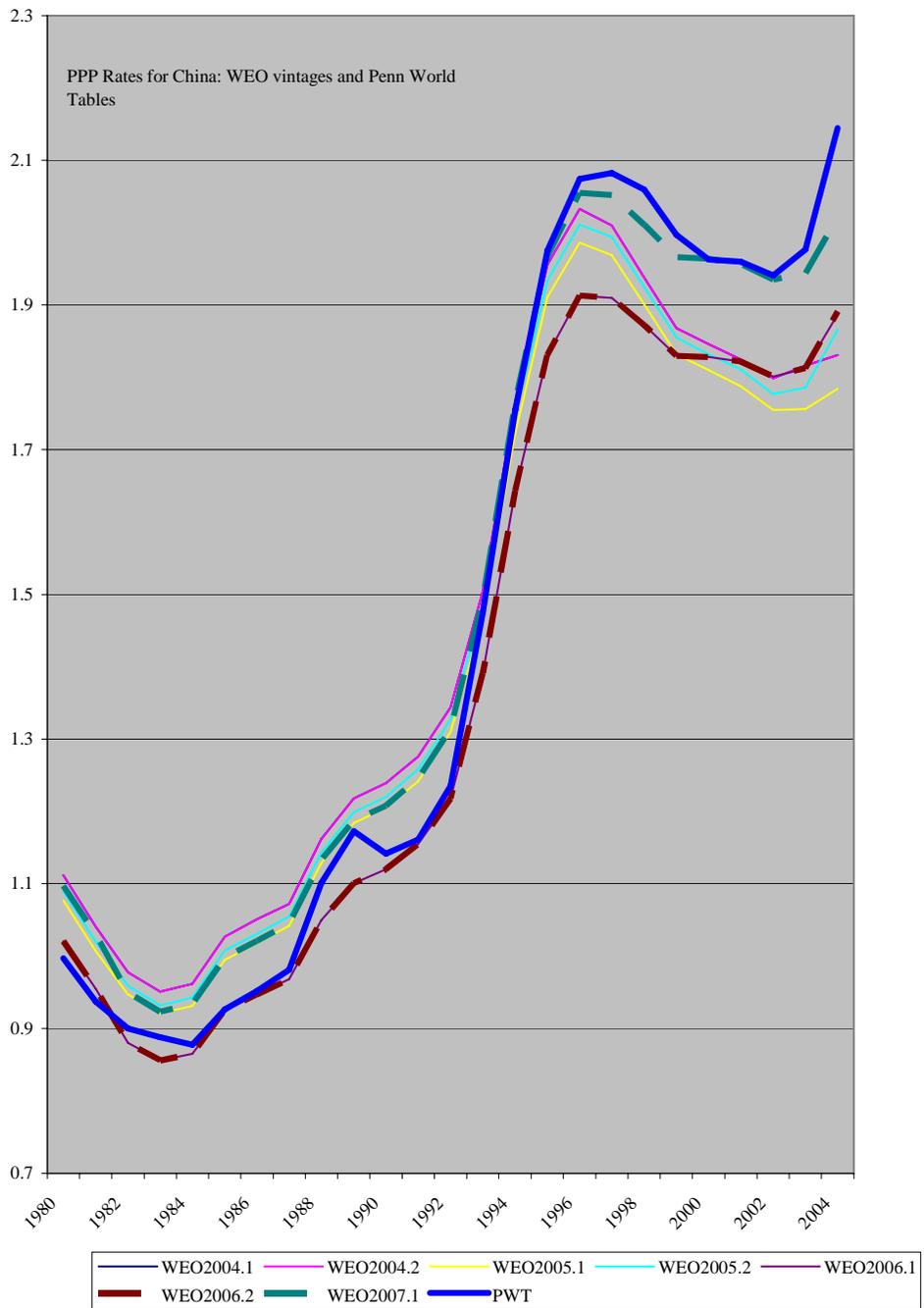


Figure 7: IMF PPP \$ Rates for China – Sensitivity to Data Vintage

reliance on Penn data overstates U.S. relative prices vis-a-vis these countries by about 20 percent. The bottom panel of figure 8 shows how these measurement errors affect WARP.<sup>20</sup> When the correction is applied only to China, the upward trend in U.S. international relative prices remains in place. Applying the correction factor to either Latin America or other emerging economies (except China and Latin American countries), leaves the trend rate of WARP largely unchanged. Finally, applying the correction factor to *all* of the emerging economies dampens the upward trend of WARP but by no means eliminates it.

### 5.2.3 Consistency with Big Mac Prices

We now evaluate whether the results from using Penn's parities are unique by comparing them to the prices of McDonald's Big Mac reported by *The Economist*. This alternative is of interest because *The Economist* reports the absolute dollar-price levels for the Big Mac. Figure 9 shows the cross-country dispersion of dollar prices for the Big Mac from 1986 to 2007 for 31 countries.<sup>21</sup> The data reveal that the number of countries with prices below the U.S. price has increased markedly over the years. As for the range of prices, Switzerland tends to have the highest price whereas China generally has the lowest price.

Given these prices, we construct the U.S. bilateral relative price of a Big Mac,  $q_{B,i}$ , as

$$q_{B,i} = \frac{P_{B,us}}{P_{B,i}},$$

where  $P_{B,us}$  is the dollar price of a Big Mac in the United States and  $P_{B,i}$  is dollar price of a Big Mac in the  $i$ th country. Given  $q_{B,i}$ , the associated geometric aggregate is

$$Q_{Bt}^g = \prod_{i=1}^{N_B} (q_{B,i})^{\omega_{B,i}}, \quad (9)$$

where  $N_B$  is the number of countries included in the aggregate and  $\omega_{B,i}$  is the trade weight for the  $i$ th country; we construct this aggregate for countries that are included in the Federal Reserve's Broad Real Dollar index.

Because the list of countries reported by the *Economist* varies across time, we construct  $Q_{Bt}^g$

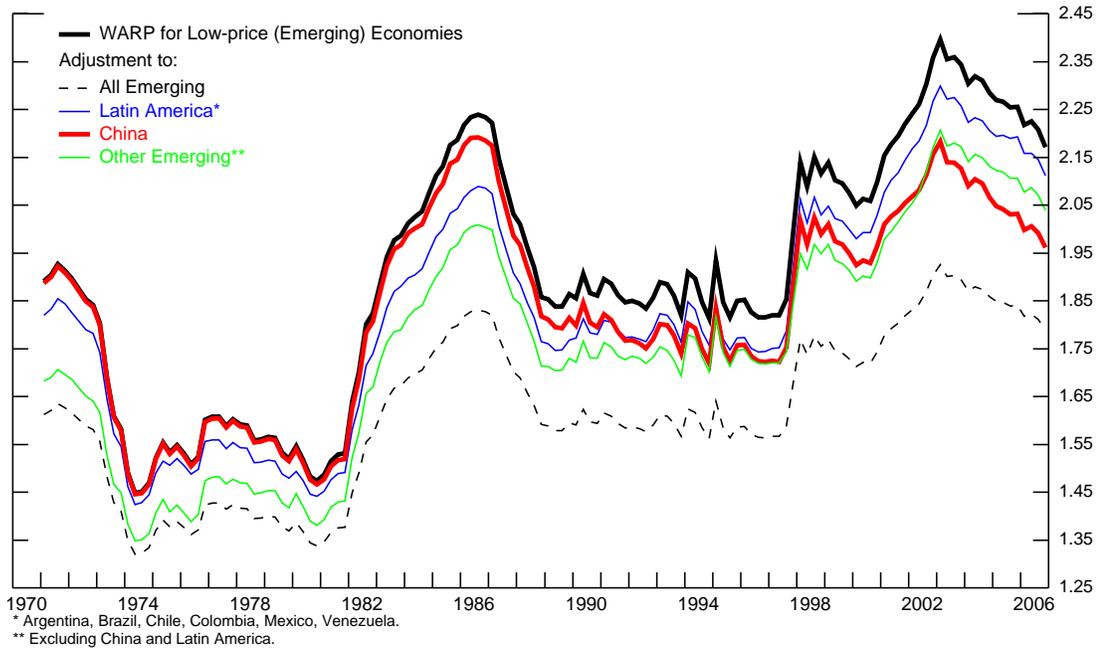
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<sup>20</sup> Assuming that the measurement error is concentrated in countries  $i = n_0 + 1 \dots n$ , the WARP is constructed as

$$\tilde{Q}_t^g = \left( \prod_{i=1}^{n_0} (q_{it})^{w_{it}} \right) \cdot \left( \prod_{i=n_0+1}^n (q_{it})^{\theta \cdot w_{it}} \right) = Q_t^g \cdot \left( \prod_{i=n_0+1}^n (q_{it})^{w_{it}} \right)^{\theta-1}.$$

<sup>21</sup> The data for this section was collected by Jeffrey Traczynski. Note that for countries that adopted the euro, *The Economist* reports prices beginning in 1999.

Geometric Aggregate for Low-price (Emerging) Economies



Geometric Aggregate for All Countries

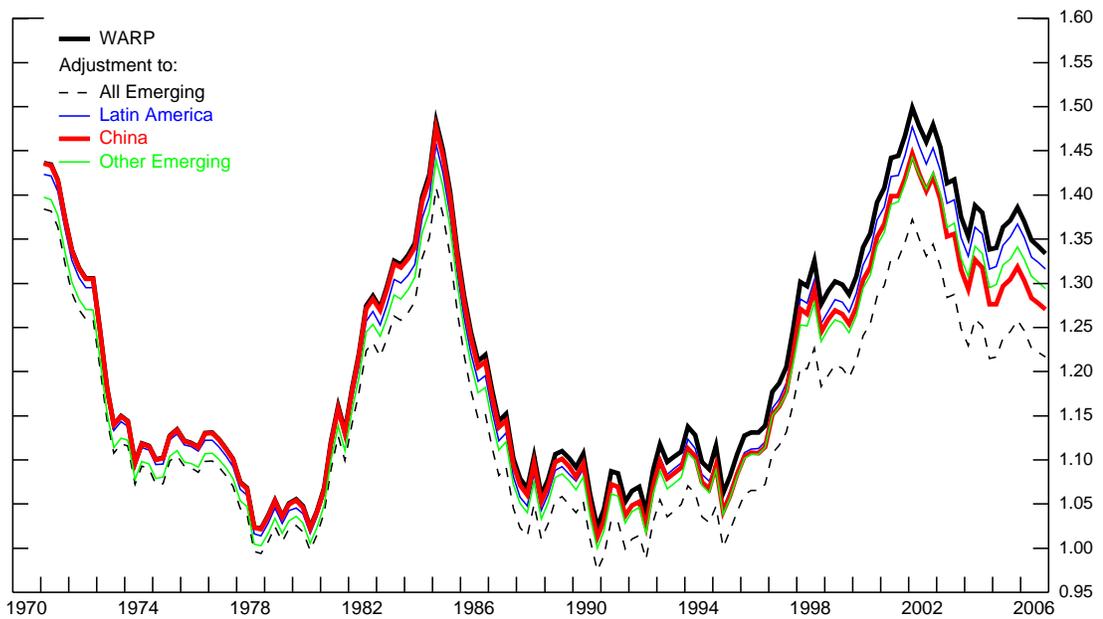


Figure 8: U.S. International Relative Prices – Sensitivity to Measurement Error

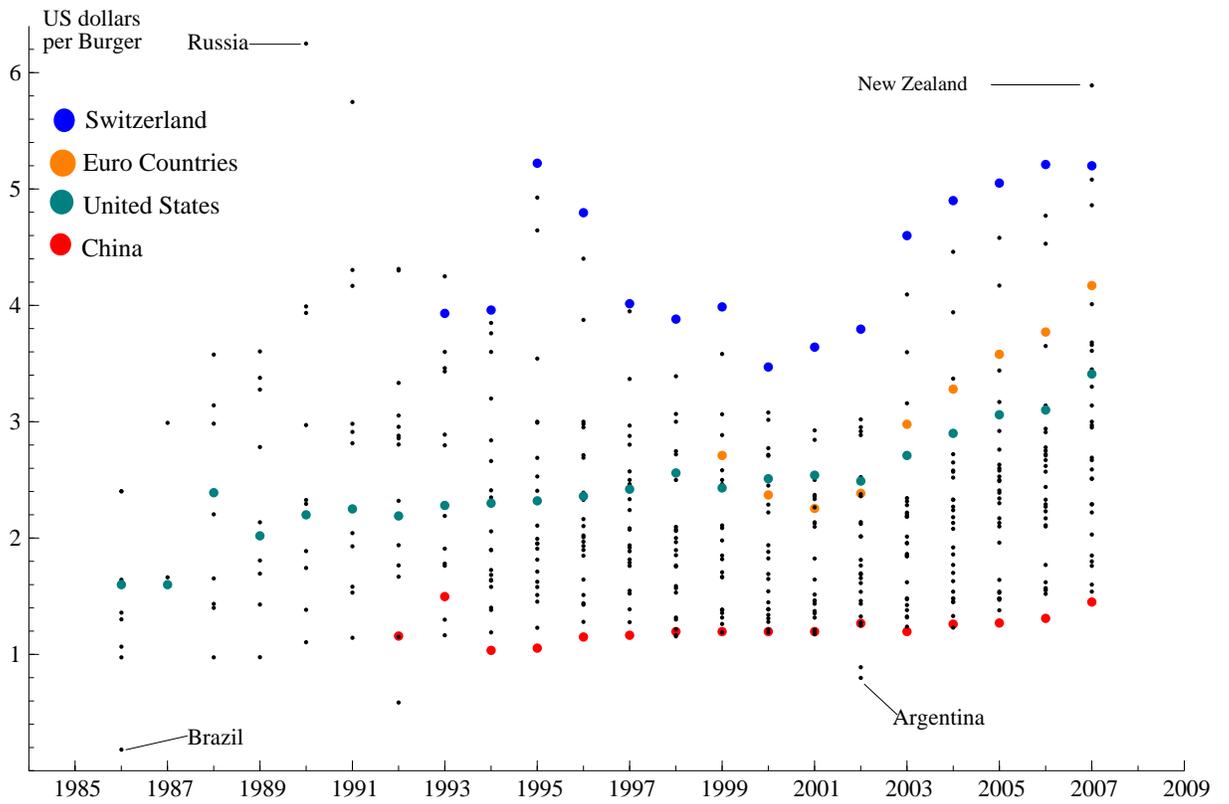


Figure 9: Cross-country Dispersion of Big Mac Prices - 1986-2007

for two groups that differ in the span of continuous data: group A with data since 1994 and group B with data since 1999; the list of countries in group B includes the countries of group A along with the euro area and other emerging economies. For comparison purposes, we also compute the chained aggregate of relative CPIs and the geometric aggregate of Penn's parities for each country group. Figure 10 shows that the aggregate of Big Mac relative prices and the WARP for group B increase from 1999 to 2006. In contrast, the chained aggregate of relative CPIs for group B declines between these two dates; comparisons based on group A give the same result. Thus we interpret these features as corroborating the evidence embodied in the WARP: U.S. international relative prices have increased.

## 6 Applications

### 6.1 WARP and Competitiveness

One question of interest is whether our WARP is informative for issues involving international competitiveness. A priori, one could argue that the WARP is not informative because it depends importantly on the prices of non-tradeables. There are, however, developments related to relative prices across countries that are not well reflected in standard measures of real exchange rates and yet have important influences on trade and other macro variables. In particular, our aim is to assess whether the WARP captures some aspects of what people have in mind when they use the term 'competitiveness' in a macro context.

It is common practice to consider competitiveness in terms of the prices of tradeable products at home and abroad:

A change in the relative price of a manufactured product (tradable good) between any two suppliers is defined as a change in price competitiveness. (McGuirk, 1986, page 3)

However, this view is somewhat sharpened by recognizing that prices of competing goods influence each other and that differences in competitiveness are determined by differences in costs that manifest themselves as differences in margins in the tradeable sector:

One might say that an industry is internationally competitive if it produces tradables and is profitable. A reduction in competitiveness is then a reduction in profitability in some or all tradables industries. (Corden 1994, page 267)

This latter view is a return to Keynes' view in 1925 when he wrote

### U.S. International Relative Prices: Big Mac versus WARP Sensitivity to Country Composition

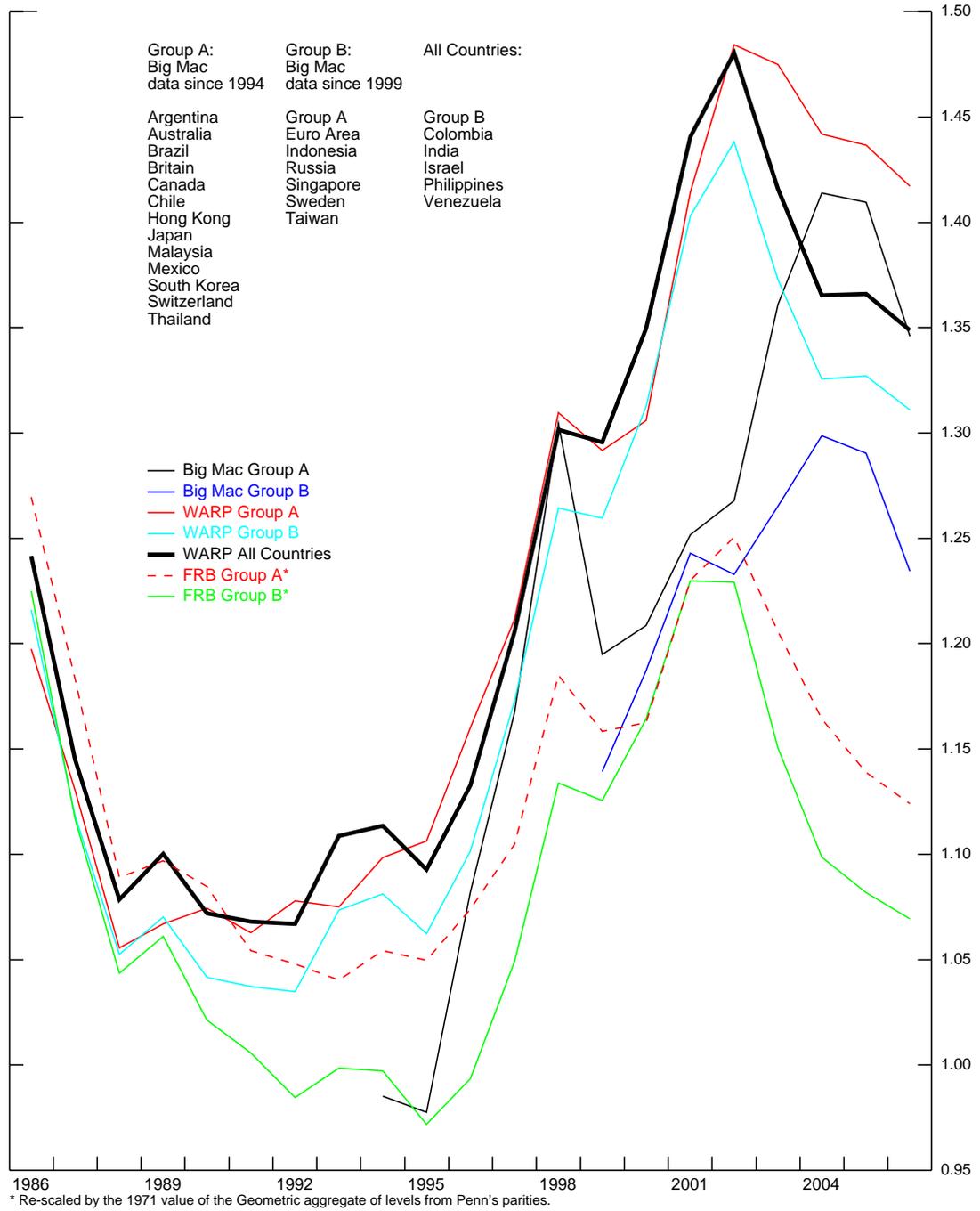


Figure 10: U.S. International Relative Prices: WARP and Big Mac

My own guess is that, compared with 1913, sheltered [non-tradeable] prices here are, at the present rates of sterling exchange, perhaps as much as ten per cent. too high in comparison with the unsheltered [tradeable] prices, and that the injury thus caused to the *competitive position* of our exports in the international market is aggravated by the fact that in Germany, France, Belgium and Italy the sheltered prices are fully ten per cent. too low. (Keynes, 1925, page 301). Emphasis and bracketed entries added.

Keynes recognized that the prices of "unsheltered" (traded) products would be nearly equalized in the world market and that competitiveness would be determined by, and reflected in, the relative prices of sheltered (non-traded) goods.

One way to illustrate Keynes' point is to show that conventional measures of competitiveness are not invariant to developments in the non-tradeable sector. Specifically, following Corden, we express competitiveness for the  $j$ th product as the ratio of producers' markups

$$\rho_j = \left( \frac{\bar{p}_j}{c_j^1} \right) / \left( \frac{\bar{p}_j}{c_j^u} \right) = \frac{c_j^u}{c_j^1},$$

where we assume that the law of one price holds with  $\bar{p}_j$  being the associated price;  $c_j^1$  is the marginal cost of the  $j$ th product in country 1 and  $c_j^u$  is the U.S. counterpart. If  $\rho_j > 1$ , then "Country 1 is said to be more competitive than the United States in the  $j$ th industry."

To determine marginal costs, appendix A.4 develops a simple, three country model in which production takes place with a Leontieff technology using labor and both tradeable and non-tradeable intermediate inputs. With these assumptions, marginal costs are linear functions of factor prices and input requirements. Thus, as detailed in equation (31) in appendix A.4, the effect on  $\rho_j$  of a one percent decrease in U.S. productivity of the non-tradeable input is

$$\frac{d\rho_j}{\rho_j} = \nu_{jn}^u > 0,$$

where  $\nu_{jn}^u$  is the share of the non-tradeable input in  $c_j^u$ . This dependency of  $\rho_j$  on non-tradeables embodies Keynes' point: questions of competitiveness cannot be usefully examined by abstracting from the prices of non-tradeables.

Overall, this example shows that if the essence of a measure of competitiveness is invariance to developments in the non-tradeable sector, then a popular measure offered in the literature is as deficient as our WARP. This difficulty, which arises independently of any compounding issues related to data availability, is simply an example of where it makes sense to tailor one's tool to address the

question at hand. Thus the suitability of a given measure to a particular application is largely an empirical question which we now examine using WARP.

## 6.2 WARP and U.S. External Imbalances

A central tenet of macroeconomic theory for open economies is that, other things equal, an increase in a country's prices relative to prices abroad will result in a deterioration in net exports. There is less agreement, however, on how to measure relative prices, and the empirical validity of the tenet clearly depends on how prices are measured. To examine this idea, figure 11 shows scatter diagrams between U.S. non-oil net exports, as a share of GDP, and the contemporaneous value of U.S. international relative prices using three measures: chained aggregate of Penn parities ( $Q_t^c$ ), chained aggregate of relative CPIs ( $Q^c$ ), and the geometric aggregate of levels of Penn parities ( $Q^g$ ). For 1971 to 2006 (left panels), the data indicate that whether net exports are inversely related to the U.S. international relative price depends on how one measures that price. Indeed, the association is absent if one uses  $Q_t^c$  where it is present if one uses  $Q^c$  and strongest if one uses  $Q^g$ .

A full understating of these disparate correlations involves recognizing that the character of bivariate associations is influenced by the level of aggregation and by omitted factors, such as foreign income and dynamic adjustments, the role of which could depend on the measure of relative prices. We will examine the role of these factors below but, in the meantime, closer inspection of the scatter plots reveals clusters of observations in which net exports and relative prices are inversely related. These clusters are most distinct for the plots using the chained aggregates of either CPIs or Penn's parities. Thus the panels on the right re-examine the relationship using the clusters formed with data from 1971-1986 and from 1987-2006; this dating is motivated by the evidence of figure 3 showing a break in the trends of international relative prices in 1987. For each of these subsamples, there is an inverse relationship between external balances and the U.S. international relative price. For the first sample, the strength of the relationship is comparable across the three measures of international relative prices. For the second sample, the association strengthens only for the geometric aggregate.

Overall, reliance on  $Q^g$  offers the greatest empirical support for the textbook proposition that net exports and relative prices are inversely related. Nevertheless, the evidence raises several questions that need to be addressed before declaring that  $Q^g$  passes the "proof of concept" test. Specifically, as indicated earlier, how can one be sure that these correlations are not unduly influenced by the absence of other key factors such as foreign economic activity?

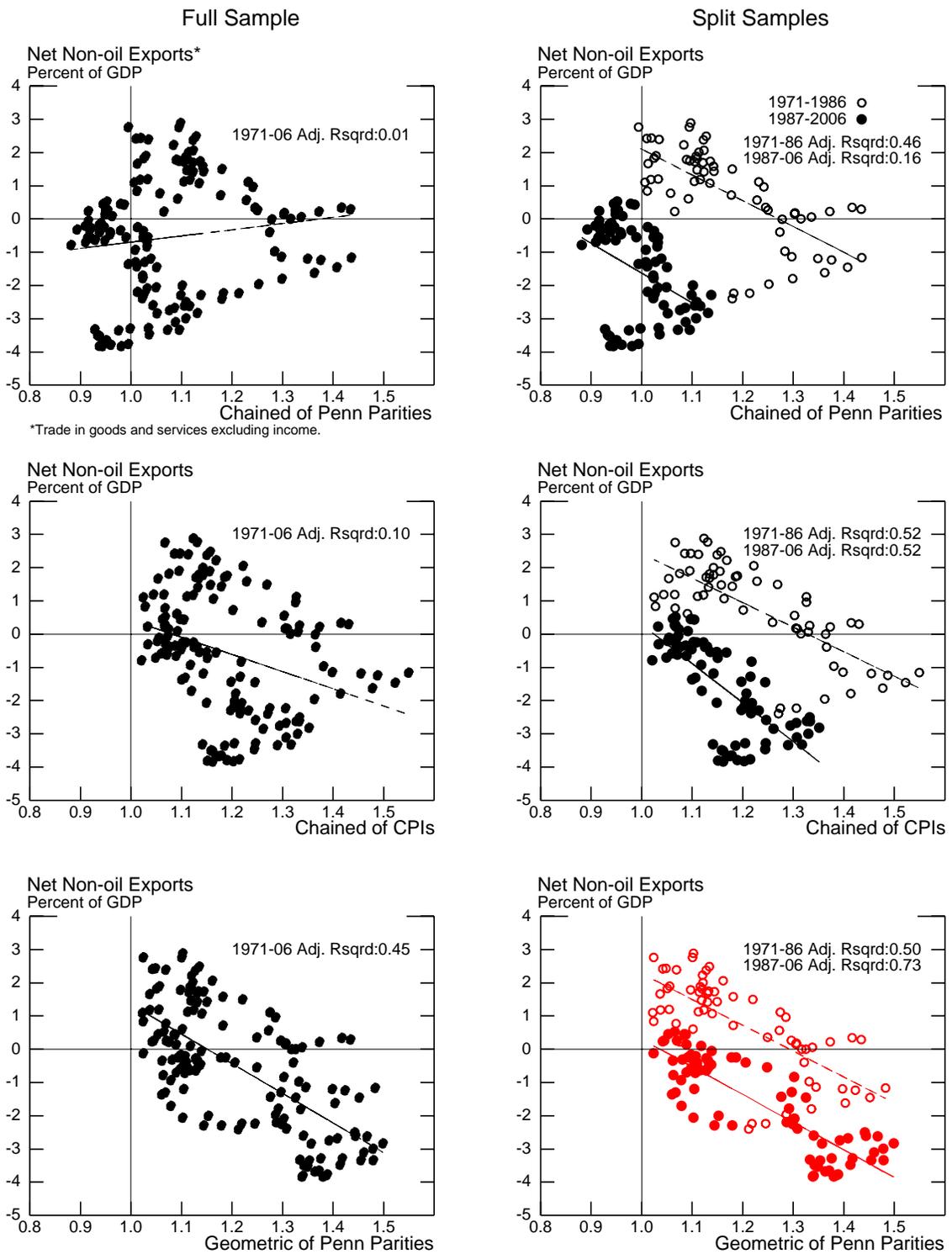


Figure 11: U.S. Non-oil Net Exports and U.S. International Relative Prices

### 6.3 WARP and Trade Modeling: The Case of Exports

To address these questions, we examine whether the alternative measures of international relative prices have implications for characterizing the behavior of aggregate U.S. exports. We focus on exports because they are directly related to foreign prices, the objective of WARP. Indeed, if one postulates that exports respond to foreign economic activity and to the price of exports ( $P_x$ ) relative to the foreign price ( $P_f$ ),  $\frac{P_x}{P_f}$ , then one needs to construct a measure for  $\frac{P_x}{P_f}$ . The current practice is to use official statistics for  $P_x$  and to measure  $P_{ft}$  as a chained aggregate of foreign CPIs, expressed in U.S. dollars. In contrast, our approach is to measure  $P_f$  so as to ensure consistency with the evolution of U.S. international relative prices. For example, if one adopts  $Q^g$  as the relevant measure of international relative prices, then  $P_f^g = \frac{P_{us}}{Q^g}$  where  $P_{us}$  is the U.S. GDP deflator.<sup>22</sup> Alternatively, if one adopts  $Q_t^c$ , then the aggregate measure of foreign prices is  $P_f^c = \frac{P_{us}}{Q^c}$ , where  $P_{us}$  is either the U.S. GDP deflator or the U.S. CPI, depending on how  $Q^c$  is constructed.

We do not focus on modeling imports because the advantage of our measure of international relative prices is less obvious given the availability of official statistics for the components of the relative price of imports.<sup>23</sup>

**Measuring the Relative Price of Exports** To measure the relative price of exports we first tailor the weighting scheme to exports and re-compute our three measures of U.S. international relative prices using weights that exclude the contribution of imports and include the role of bilateral exports and third-country markets; figure 12 shows that the choice of weights has a relatively minor effect on our three measures of international relative prices. Second, we solve for the implied  $P_f$  :

$$P_{fx}^g = \frac{P_{us}}{Q_x^g} : \text{geometric of levels of Penn parities} \quad (10)$$

$$P_{fx}^c = \frac{P_{us}}{Q_x^c} : \text{chained of levels of Penn parities}$$

$$CPI_{fx}^c = \frac{CPI_{us}}{Q_x^c} : \text{chained of CPIs,}$$

<sup>22</sup>Note that we back out the foreign price with the standard U.S. GDP deflator which does not have, exactly, the same basket as that used by the Penn World Tables.

<sup>23</sup>Preliminary work (available on request) indicates that  $Q^g$  helps in correcting the estimation bias associated with how new products are measured in official import prices,  $P_m$ ; see Hooper and Richardson (1991) and Feenstra (1994). Furthermore, our measures of  $P_f$  are relevant for explaining  $P_m$  which is normally modeled as a function of foreign prices; see Thomas and Marquez (2006) for further details on the modeling of  $P_m$ .

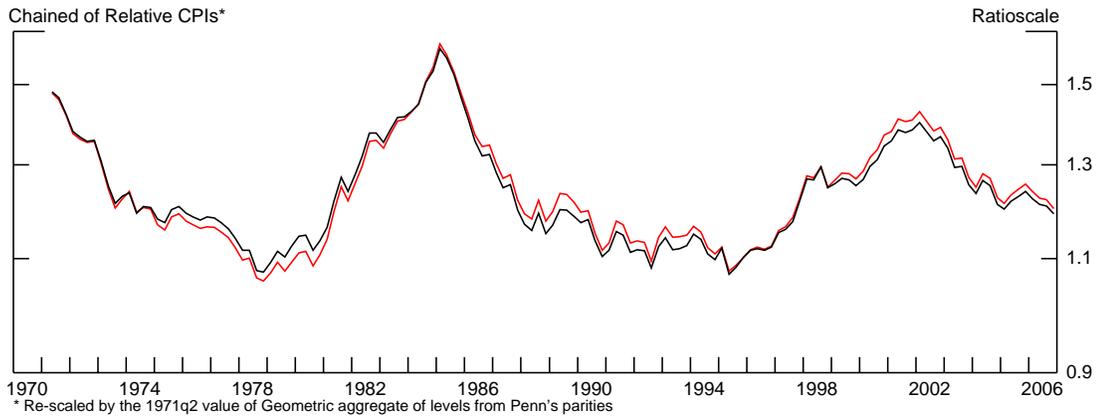
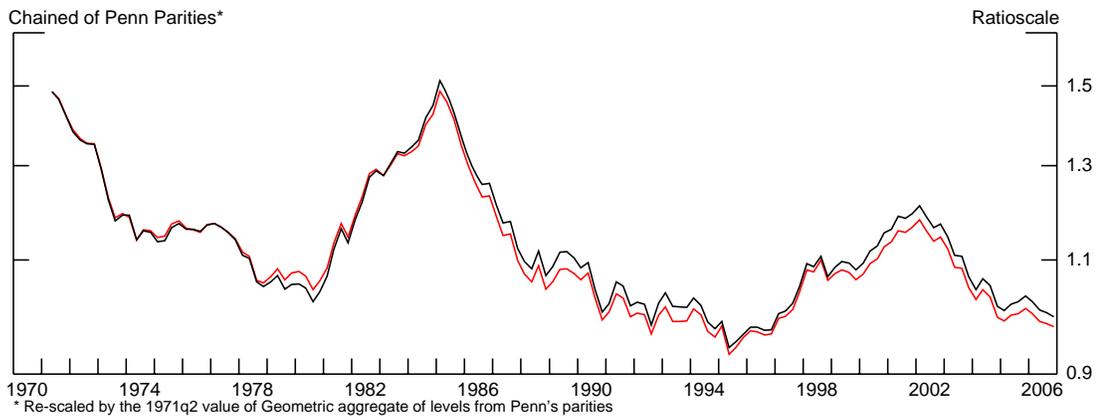
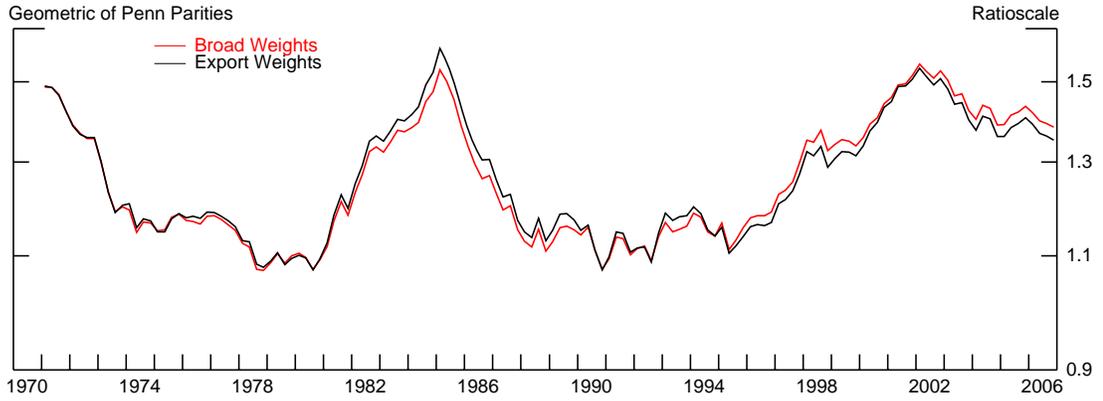


Figure 12: U.S. International Relative Prices - Sensitivity to Weighting Scheme

where a sub-script ‘ $x$ ’ denotes the use of export weights and  $CPI_{fx}^c$  is the chained aggregate of foreign CPIs, expressed in U.S. dollars. Third, given  $P_x$ , we obtain the three measures of relative export prices as

$$rpx^{geo} = \frac{P_x}{P_{fx}^g} \quad (11)$$

$$rpx^c = \frac{P_x}{P_{fx}^c}$$

$$rpx^{cpi} = \frac{P_x}{CPI_{fx}^c}.$$

How important are differences in international relative prices for the profile of the relative price of exports? Figure 13 documents the data and the steps taken to arrive at the three measures of  $rpx$ . The top-left panel shows the (export-weighted) measures of international relative prices; the right panel shows the three U.S. price indexes:  $P_{us}$ ,  $CPI_{us}$ , and  $P_x$ . The bottom-left panel shows the implied measures of aggregate foreign prices (equation 10); the series have upward trends and move together through 1987 but diverge afterwards with  $P_{fx}^g$  flattening while the other two series continue their upward trends, albeit at a lower rate. The flattening of  $P_{fx}^g$  reflects the increasing importance of low-price economies, a phenomenon captured only by  $Q_x^g$ . The bottom-right panel shows the three measures for  $rpx$  (equation 11);  $rpx_t^c$  and  $rpx_t^{cpi}$  move together and have downward trends reflecting the upward trend in their measures of foreign prices. In contrast,  $rpx_t^{geo}$  trends down through 1990 and flattens afterwards, reflecting the flattening of  $P_{fx}^g$ . Overall, aggregation schemes that recognize interactions between price levels and the increased trade with emerging economies, as captured by  $Q_x^g$ , yield a picture of U.S. relative export prices that is fundamentally different from the one given by existing aggregation methods. We now examine whether this difference matters for characterizing the response of U.S. exports to income and relative prices.

**Econometric Formulation** To model U.S. exports, we assume that foreign and domestic products are imperfect substitutes for each other (see Goldstein and Khan, 1985) and postulate an

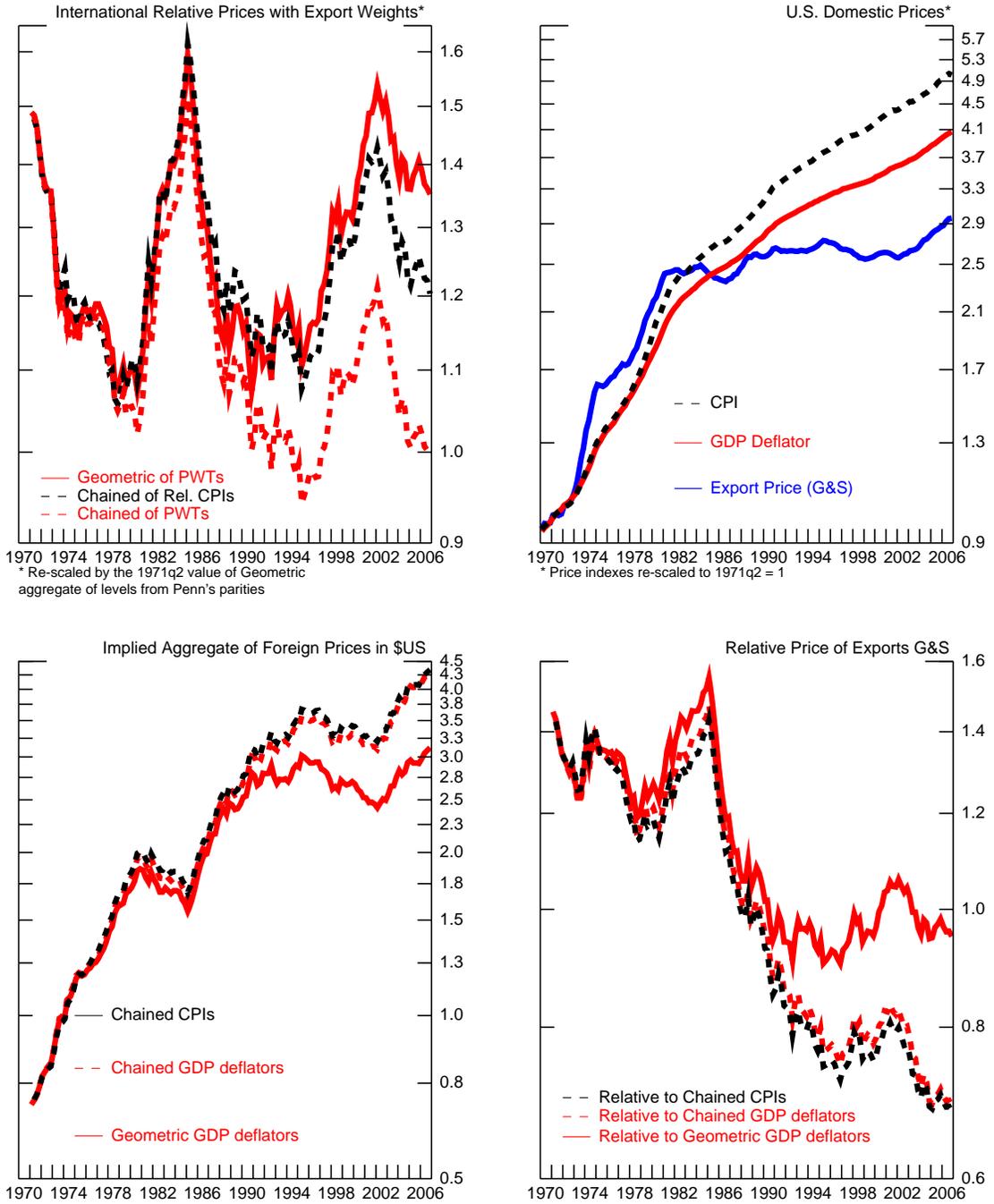


Figure 13: Derivation of U.S. Relative Export Price

error-correction formulation:

$$\begin{aligned} \Delta \ln X_t &= \underbrace{\alpha^j + \sum_{i=1}^4 \beta_i^j \cdot \Delta \ln X_{t-i} + \sum_{i=0}^4 \phi_i^j \cdot \Delta \ln Y_{t-i}^* + \sum_{i=0}^4 \mu_i^j \cdot \Delta \ln rpx_{t-i}^j}_{short-run} + \underbrace{\theta_x^j \cdot (\ln X_{t-1} - \eta_x^j \cdot \ln Y_{t-1}^* - \varepsilon_x^j \cdot \ln rpx_{t-1}^j)}_{long-run} + u_t^j, \quad u_t^j \sim IN(0, \sigma_j^2) \\ j &= geo, c, cpi, \end{aligned} \quad (12)$$

where  $X$  is the volume of exports of goods and services;  $Y^*$  is the foreign real GDP;  $\eta_x^j > 0$  is the long-run income elasticity; and  $\varepsilon_x^j < 0$  is the long-run price elasticity.<sup>24</sup> Equation (12) assumes that the growth rate of exports responds to short- and long-run factors. Specifically, movements in income and relative prices induce cyclical swings in exports. But, even if income and relative prices were fixed, exports could be changing as they adjust to their long run level given by  $\eta_x^j \cdot \ln Y^* + \varepsilon_x^j \cdot \ln rpx^j$ . This gradual adjustment is captured by the term in parentheses where  $\theta_x^j < 0$  represents the speed of adjustment. Finally, finding that  $\alpha^j \neq 0$  means that exports would automatically change over time regardless of the evolution of income and relative prices and thus we interpret a significant  $\alpha^j$  as evidence of misspecification.

As formulated, equation (12) is non-linear in the parameters. To avoid the associated estimation difficulties, we re-express this equation as linear in the parameters:

$$\begin{aligned} \Delta \ln X_t &= \alpha^j + \sum_{i=1}^4 \beta_i^j \cdot \Delta \ln X_{t-i} + \sum_{i=0}^4 \phi_i^j \cdot \Delta \ln Y_{t-i}^* + \sum_{i=0}^4 \mu_i^j \cdot \Delta \ln rpx_{t-i}^j + \theta_x^j \cdot \ln X_{t-1} + \theta_y^j \cdot \ln Y_{t-1}^* + \theta_p^j \cdot \ln rpx_{t-1}^j + u_t^j, \quad u_t^j \sim IN(0, \sigma_j^2) \\ j &= geo, c, cpi; \theta_y^j > 0 \text{ and } \theta_p^j < 0. \end{aligned} \quad (13)$$

Using a ‘ $\hat{\phantom{x}}$ ’ to denote an estimated value, we use the least squares values of  $\hat{\theta}_x^j$ ,  $\hat{\theta}_y^j$ , and  $\hat{\theta}_p^j$  to compute the implied elasticities as  $\hat{\eta}_x^j = -\frac{\hat{\theta}_y^j}{\hat{\theta}_x^j}$  and  $\hat{\varepsilon}_x^j = -\frac{\hat{\theta}_p^j}{\hat{\theta}_x^j}$ . Note that these elasticity estimates are ratios of normal variables and thus the associated distributions are not known in advance.<sup>25</sup> Thus the associated confidence intervals are constructed using Monte Carlo simulations; appendix A.5 has the details.

<sup>24</sup>Based on the results from the Augmented Dickey-Fuller test, one cannot reject a unit-root for the level of exports, foreign income, and the three measures of relative prices. See table A.1 in the data appendix.

<sup>25</sup>See Marsaglia (1965) and Anderson and Thursby (1986) for details.

**Estimation Strategy** For parameter estimation we apply least squares to equation (13) using observations from 1972Q3 to 2004Q4 with data from 1971Q2 to 1972Q2 reserved for lags and data from 2005Q1 to 2006Q4 reserved for evaluating out-of-sample predictive accuracy. One may argue that there are gains in precision of the estimates if one were to exclude insignificant variables from the model. To avoid the statistical pitfalls associated with the joint nature of model specification and parameter estimation, we rely on a computer-automated algorithm, developed by Hendry and Krolzig (2001).<sup>26</sup> Their algorithm combines least squares with a selection criteria that excludes insignificant coefficients and tests for both parameter constancy and white-noise residuals; the critical values for rejection are not fixed in advance but, rather, are calculated sequentially. We report results for equation (13), labeled the General formulation, and for the simplified formulation, labeled the Specific formulation.

To examine the potential for simultaneity bias, we postulate a vector-autoregressive model explaining exports, income, and relative prices and then apply Johansen’s cointegration method to estimate the cointegration vector; this approach treats income and prices as endogenous.<sup>27</sup>

**Econometric Results** Table 1 shows estimation and test results for all three measures of relative export prices. The signs for  $\theta_x^j$ ,  $\theta_y^j$ , and  $\theta_p^j$  are consistent with expectations and their magnitudes are roughly comparable across measures of relative export prices. In terms of in-sample fit, the standard error of the regression has a narrow range of variation: from 1.89% for  $rp x_t^{geo}$  to 1.93% for  $rp x_t^{cpi}$ . Furthermore, the Chow tests cannot reject the hypothesis of parameter stability, and the residuals exhibit normality, serial independence, and homoskedasticity.

The sole dissonant note in these results is the presence of a positive and statistically significant intercept in the specific formulation using the chained of relative CPIs (column 6). Finding that  $\alpha^{cpi} > 0$  means that exports would expand even if income and relative prices were fixed. We do not see an economic justification for such a result and treat this finding as an instance in which an algorithm delivering an otherwise statistically reliable model is not delivering an economically meaningful model. Thus we also re-estimate the parameters of the model constraining the intercept to zero and find (column 7) that the constrained model exhibits a slight deterioration of fit, which is not surprising, and that the values for the remaining parameter estimates based on  $rp x_t^{cpi}$  are close to the estimates based on  $rp x_t^c$ . (This finding is reassuring given the similarity in the data for these

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<sup>26</sup>For a discussion of the issues raised by automated specification, see Hendry and Krolzig (2003), Granger and Hendry (2004), and Phillips (2004).

<sup>27</sup>Results from the Johansen method are sensitive to the number of lags included in the VAR. To recognize this feature, we estimate VARs with alternative lags: from 12 quarters to 3 quarters. The estimates reported here correspond to the number of lags that maximizes the probability of having one cointegrating vector.

Table 1: Parameter Estimates for Model of U.S. Exports of Goods and Services: Sensitivity to Measure of Foreign Prices-OLS, 1971Q2-2004Q4

|   | Geometric Foreign PGDP |                 | Chained Foreign PGDP |                 | Chained Foreign CPIs |                 |                      |
|---|------------------------|-----------------|----------------------|-----------------|----------------------|-----------------|----------------------|
|   | General<br>(1)         | Specific<br>(2) | General<br>(3)       | Specific<br>(4) | General<br>(5)       | Specific<br>(6) | ex. Intercept<br>(7) |
| <b>Parameter Estimates</b>              |                        |                 |                      |                 |                      |                 |                      |
| Intercept $\alpha$                      | 0.0392                 | 0e              | 0.1618               | 0e              | 0.1966               | 0.1574          | set to zero          |
| se                                      | 0.0474                 |                 | 0.0664               |                 | 0.0725               | 0.0657          |                      |
| Sum of coefficients for $\Delta \ln X$  | -0.3157                | 0e              | -0.1478              | 0e              | -0.1448              | 0e              | 0e                   |
| se                                      | 0.192                  |                 | 0.186                |                 | 0.1829               |                 |                      |
| Sum of coefficients for $\Delta \ln Y$  | 4.3777                 | 2.899           | 3.8885               | 3.1748          | 4.0375               | 2.8246          | 3.2633               |
| se                                      | 0.79                   | 0.4267          | 0.7885               | 0.4247          | 0.786                | 0.4571          | 0.4267               |
| Sum of coefficients for $\Delta \ln RP$ | 0.088                  | 0e              | 0.1178               | 0e              | 0.1661               | 0e              | 0e                   |
| se                                      | 0.1885                 |                 | 0.2109               |                 | 0.1954               |                 |                      |
| Lagged Exports: $\theta_x$              | -0.1244                | -0.1157         | -0.1547              | -0.1249         | -0.1534              | -0.1295         | -0.1138              |
| se                                      | 0.0368                 | 0.0260          | 0.0448               | 0.0299          | 0.0404               | 0.0292          | 0.0289               |
| Lagged Foreign Income: $\theta_y$       | 0.1735                 | 0.1709          | 0.1860               | 0.1813          | 0.1748               | 0.1508          | 0.1644               |
| se                                      | 0.0579                 | 0.0391          | 0.0626               | 0.0443          | 0.0552               | 0.0423          | 0.0427               |
| Lagged Relative Price: $\theta_p$       | -0.1314                | -0.0982         | -0.1439              | -0.0755         | -0.1503              | -0.1195         | -0.0675              |
| se                                      | 0.0315                 | 0.0203          | 0.0368               | 0.0171          | 0.0358               | 0.0269          | 0.0162               |
| <b>Measures of Fit</b>                  |                        |                 |                      |                 |                      |                 |                      |
| SER                                     | 1.807%                 | 1.890%          | 1.812%               | 1.917%          | 1.809%               | 1.895%          | 1.930%               |
| Adj Rqrd                                | 0.459                  | 0.408           | 0.455                | 0.391           | 0.457                | 0.405           | 0.382                |
| No. of Parameters                       | 18                     | 5               | 18                   | 5               | 18                   | 6               | 5                    |
| <b>Hypotheses (p-values) (a)</b>        |                        |                 |                      |                 |                      |                 |                      |
| <b>Parameter Stability</b>              |                        |                 |                      |                 |                      |                 |                      |
| Half Sample                             | 0.999                  | 0.998           | 0.996                | 0.996           | 0.998                | 0.995           | 0.998                |
| Last 8 quarters                         | 0.973                  | 0.858           | 0.835                | 0.589           | 0.908                | 0.788           | 0.700                |
| <b>Properties of residuals</b>          |                        |                 |                      |                 |                      |                 |                      |
| Normality                               | 0.020                  | 0.108           | 0.039                | 0.175           | 0.026                | 0.153           | 0.170                |
| Serial Independence                     | 0.841                  | 0.093           | 0.578                | 0.417           | 0.174                | 0.174           | 0.299                |
| Homoskedasticity                        | <b>0.009</b>           | 0.137           | 0.016                | 0.188           | 0.150                | 0.150           | 0.140                |
| <b>Implied Elasticities</b>             |                        |                 |                      |                 |                      |                 |                      |
| Income = $-\theta_y / \theta_x$         | 1.39                   | 1.48            | 1.20                 | 1.45            | 1.14                 | 1.16            | 1.44                 |
| Price = $-\theta_p / \theta_x$          | -1.06                  | -0.85           | -0.93                | -0.60           | -0.98                | -0.92           | -0.59                |

0e: Algorithm finds the variable to be statistically irrelevant and sets the coefficient to zero.

(a): An entry less than 0.01 means that the associated hypothesis can be rejected at the 1% significance level.

two relative prices.) In terms of the elasticities, the implied income elasticity is positive and ranges from 1.4 for  $rp x_t^{cpi}$  to 1.5 for  $rp x_t^{geo}$ ; the implied price elasticity is negative and ranges from -0.6 for  $rp x_t^{cpi}$  to -0.9 for  $rp x_t^{geo}$ .

To assess the statistical properties of these estimates, figure 14 shows the 95% (Monte Carlo) confidence intervals for the estimated income elasticity along with the confidence bands for estimates from the Johansen method; figure 18 in appendix A.5 shows the densities for all income elasticities. The results indicate that the median income elasticity is positive; greater than one; significantly greater than zero; and quite similar to the implied income elasticity of table 1. Furthermore, the median elasticity based on  $rp x_t^{geo}$  exceeds the median elasticity for the other measures of relative prices, a result robust to estimation method. Note that the proximity of the median to the 95% bound means that the empirical distribution is not symmetrical. Finally, the estimates from the Johansen method are quite close to the estimates from the General formulation; this finding suggests that simultaneity biases, if present, are not affecting the income elasticity.

Figure 15 reports the 95% (Monte Carlo) confidence intervals for the estimated price elasticity; figure 19 in appendix A.5 shows the densities for all price elasticities. The results indicate that the median price elasticity is negative, significantly below zero, and quite close to the implied price elasticity of table 1. Furthermore, the median elasticity based on  $rp x_t^{geo}$  exceeds (in absolute value) the median elasticity for the other measures of relative prices, a result robust to estimation method. Finally, the estimates from the Johansen method are quite close to the estimates from the General formulation.

We find that, unless one has strong priors about the values of income and price elasticities, the results do not allow us to select one measure of relative prices over another. To that end, we assess whether forecast accuracy is sensitive to the measure of relative prices. Specifically, we generate one-step ahead predictions for the growth rate of exports from 2005Q1 to 2006Q4. Table 2 reports that the root mean squared error for the specific formulations varies from 0.60 percent for  $rp x_t^{geo}$  to 1.25% for  $rp x_t^c$ ; the mean forecast error varies from -0.14 for  $rp x_t^{geo}$  to -1.12% for  $rp x_t^c$ . Overall, the formulation using  $rp x_t^{geo}$  has the lowest mean forecast error and the lowest RMSE.

Finally, we implement an encompassing test to establish which measure of relative prices offers the best statistical characterization of U.S. exports. The approach involves two steps. First, we

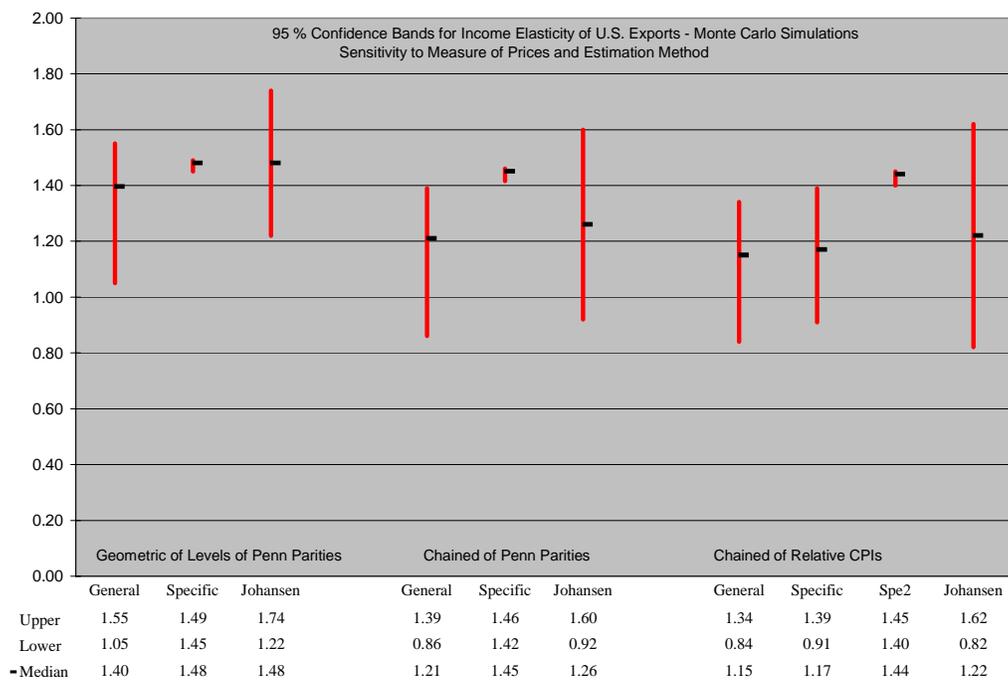


Figure 14: Long-run Income Elasticity for U.S. Exports

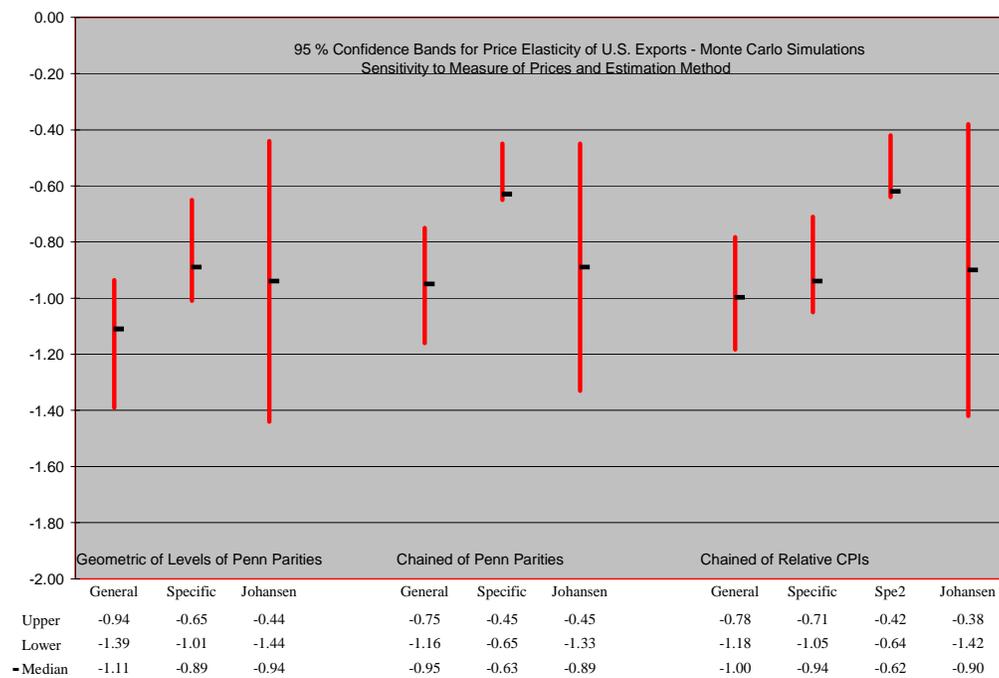


Figure 15: Long-run Price Elasticity for U.S. Exports

Table 2: Forecast Accuracy for U.S. Exports of Goods and Services: Alternative Measures, 2005Q1-2006Q4

|                         | <u>Geometric Foreign PGDP</u> |                 | <u>Chained Foreign PGDP</u> |                 | <u>Chained Foreign CPIs</u> |                 |                      |
|-------------------------|-------------------------------|-----------------|-----------------------------|-----------------|-----------------------------|-----------------|----------------------|
|                         | General<br>(1)                | Specific<br>(2) | General<br>(3)              | Specific<br>(4) | General<br>(5)              | Specific<br>(6) | ex. Intercept<br>(7) |
| Root Mean Squared Error | 0.558%                        | 0.595%          | 1.232%                      | 1.248%          | 0.889%                      | 0.749%          | 1.004%               |
| Mean Forecast Error (a) | 0.033%                        | -0.140%         | -1.098%                     | -1.123%         | -0.695%                     | -0.461%         | -0.844%              |

(a) Actual minus predicted

postulate a general model that encompasses the three formulations examined so far as special cases:

$$\begin{aligned} \Delta \ln X_t = & \alpha + \sum_{i=1}^4 \beta_i \cdot \Delta \ln X_{t-i} + \sum_{i=0}^4 \phi_i \cdot \Delta \ln Y_{t-i}^* + \theta_x \cdot \ln X_{t-1} + \theta_y \cdot \ln Y_{t-1}^* \\ & + \sum_{i=0}^4 \mu_i^{geo} \cdot \Delta \ln rpx_{t-i}^{geo} + \sum_{i=0}^4 \mu_i^c \cdot \Delta \ln rpx_{t-i}^c + \sum_{i=0}^4 \mu_i^{cpi} \cdot \Delta \ln rpx_{t-i}^{cpi} \\ & + \theta_p^{geo} \cdot \ln rpx_{t-1}^{geo} + \theta_p^c \cdot \ln rpx_{t-1}^c + \theta_p^{cpi} \cdot \ln rpx_{t-1}^{cpi} + u_t, \quad u_t \sim IN(0, \sigma^2). \end{aligned}$$

Second, we apply the automated specification algorithm to estimate the parameters of the above model. The possible outcomes are

- only one measure of relative prices matters and all others are redundant;
- one needs more than one measure of relative prices to explain U.S. exports;
- relative prices do not matter for explaining U.S. exports.

Table 3 reports the estimation results for both the general and the specific formulations. We find that reliance on  $rpx_t^{geo}$  makes redundant the alternative measures of relative prices. In other words, the geometric measure of relative export prices encompasses the other two measures of relative prices. Further, the resulting specific formulation of table 3 is the same as the specific formulation for the model using  $rpx_t^{geo}$  of table 1 (column 2).

This finding is of interest for two reasons. First, the measure of relative prices that has the strongest association with net exports in figure 11 above is also the measure of relative prices that offers the best explanation of exports. Second, the choice of measure of relative prices matters for characterizing the response of U.S. exports to changes in income and prices. Specifically, the estimated income elasticity based on  $rpx_t^{geo}$  is higher than estimates based on other measures of relative prices and higher than the unitary elasticity previously reported in the literature.<sup>28</sup> This finding is potentially relevant for addressing general questions involving the sustainability of the U.S. current account and more pointedly whether the Houthakker-Magee asymmetry of income elasticities is being influenced by the measure of relative prices.

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<sup>28</sup>See Houthakker and Magee (1969), Mann (1999), and Hooper, Johnson, and Marquez (2000).

Table 3: Encompassing Test for Model of U.S. Exports - 1971Q2-2004Q4

|  | General<br>(1) | Specific<br>(2) |
|--|----------------|-----------------|
| Intercept $\alpha$                                 | 0.19066        | 0e              |
| se   | 0.11642        |                 |
| Sum of coefficients for $\Delta \ln X$             | -0.4146        | 0e              |
| se   | 0.2261         |                 |
| Sum of coefficients for $\Delta \ln Y$             | 4.1348         | 2.899           |
| se   | 0.8351         | 0.4267          |
| Sum of coefficients for $\Delta \ln RP\text{-Geo}$ | -2.1863        | 0e              |
| se   | 0.9996         |                 |
| Sum of coefficients for $\Delta \ln RP\text{-c}$   | -0.0426        | 0e              |
| se   | 1.6391         |                 |
| Sum of coefficients for $\Delta \ln RP\text{-CPI}$ | 2.4899         | 0e              |
| se   | 1.2075         |                 |
| Lagged Exports: $\theta_x$                         | -0.1493        | -0.116          |
| se   | 0.0538         | 0.026           |
| Lagged Foreign Income: $\theta_y$                  | 0.17535        | 0.1709          |
| se   | 0.07861        | 0.0391          |
| Lagged Relative Price: $\theta_p$                  |                |                 |
| Geometric of Levels of Penn Parities               | -0.1113        | -0.0982         |
| se   | 0.07789        | 0.0203          |
| Chained of Levels of Penn Parities                 | 0.04414        | 0e              |
| se   | 0.19017        |                 |
| Chained of Relative CPIs                           | -0.1245        | 0e              |
| se   | 0.2085         |                 |
| Measures of Fit                                    |                |                 |
| SER  | 0.01771        | 0.0189          |
| Adj. $R^2$   | 0.47991        | 0.4076          |
| No. of Parameters                                  | 30             | 5               |
| <b>Hypotheses (p-values) (a)</b>                   |                |                 |
| Parameter Stability                                |                |                 |
| Half Sample  | 0.9878         | 0.998           |
| Last 8 quarters                                    | 0.9593         | 0.858           |
| Properties of residuals                            |                |                 |
| Normality  | 0.2516         | 0.108           |
| Serial Independence                                | 0.1003         | 0.093           |
| Homoskedasticity                                   | 0.1123         | 0.137           |

0e: Algorithm finds the variable to be statistically irrelevant and sets the coefficient to zero.

(a): An entry less than 0.01 means that the associated hypothesis can be rejected at the 1% significance level.

## 7 Conclusions

This paper identifies an aspect of international price developments—the interaction of differences in price levels with changing trade shares—that is not captured in conventional real exchange rate indexes and constructs a new weighted average relative price (WARP) to capture this interaction. The WARP indicates that over the past 20 years there has been a secular rise in U.S. prices relative to prices in the rest of the world when these prices are weighted by U.S. trade shares. This is in sharp contrast to what conventional measures indicate—that there has been no change in U.S. prices relative to the rest of the world. We use WARP to take a fresh look at an ongoing puzzle: how could U.S. prices relative to the rest of the world show no trend and yet U.S. net exports, as a share of GDP, have declined? We find that our WARP shows the strongest inverse association between net exports and relative prices. In other words, WARP restores the usefulness of that theoretical prediction. To examine whether this resolution is the result of a statistical artifact, we characterize the response of U.S. exports to income and relative prices and examine its sensitivity to the measure of relative price of exports used. Unless one has strong priors about the values of income and price elasticities, the results do not allow us to select one measure of relative prices over another. Thus we examine the forecast accuracy of the different models and find that the formulation based on WARP has the lowest forecast errors. Furthermore, encompassing tests indicate that the inclusion of the WARP-based measure of export's relative prices in the specification makes redundant the alternative measures of relative prices.

# A Appendixes

## A.1 Data for Bilateral Relative Prices

**Extending Parities from Penn World Tables Through 2006** For most countries, the annual data for bilateral relative prices from the Penn World Tables end in 2004. We extrapolate the data through 2006 by assuming that the Penn purchasing power parities grow from 2004 to 2006 at the same rate as the ratio of U.S. to foreign-currency GDP deflators. For Brazil, Colombia, India, Malaysia, Thailand, and Russia, the extrapolation process starts in 2003 because that is when the data end for these countries. The rationale for using the GDP deflator series in the extrapolation is that they use a GDP basket of goods similar, in theory, to the basket used by the Penn parities. The data for GDP deflators are available from the International Monetary Fund’s International Financial Statistics (IFS) database and HAVER databases.

**Quarterly Parities from Penn World Tables** The procedure uses an annual “target” series,  $T_t^1$ , and a quarterly “pattern” series,  $P_t^4$ , where the superscripts 1 and 4 refer to the frequency, either annual or quarterly. The annual target series is the series that we wish to have on a quarterly basis; the quarterly pattern series is used to guide the interpolation between values of the annual target series. For the annual target series of the  $i$ th country, we use the annual bilateral relative price from the Penn World Tables—namely,  $T_{it}^1 = q_{it}$ . For the quarterly pattern series, we use the CPI-adjusted bilateral exchange rate corrected for the systematic gap in inflation rates between the U.S. CPI and the U.S. GDP deflator:

$$P_{i,ts}^4 = \tilde{r}_{i,ts}, \quad s = 1 \dots 4,$$

where

$$\tilde{r}_{i,ts} = \left( \frac{CPI_{us,ts}}{CPI_{i,ts}} \right) \cdot \left( \frac{E_{\frac{i}{s},ts}}{E_{\frac{i}{s},t_0}} \right) \cdot \left( \frac{P_{us,ts}}{CPI_{us,ts}} \right) = r_{i,ts} \cdot \left( \frac{P_{us,ts}}{CPI_{us,ts}} \right).$$

Our choice of  $\tilde{r}_{i,ts}$  is based on two considerations. First, quarterly GDP deflators (the ideal series) are not available for several emerging economies. Second, one avoids the biases induced by the well-known wedge between the trend growth rate of the U.S. CPI and the trend growth rate of the U.S. GDP deflator; such a wedge is minimal for other countries. This wedge is relevant because the bilateral relative prices from the Penn World Tables are based on GDP prices and thus, ignoring this wedge, would bias the quarterly growth rates.

We want to emphasize that we are not relying on a quarterly bilateral exchange-rate *index* to serve as a suitable proxy to estimate quarterly data for the *level* of the bilateral relative price.

Rather, we use  $\tilde{r}_{i,ts}$  to obtain the quarterly pattern within a year of the growth rates of the bilateral relative prices. With these considerations in mind, there are 10 steps to construct quarterly data:

1. Estimate quarterly weights using a cubic spline subject to two constraints: (1) the average of the quarterly weights for a given year be the same as the annual weight for that year and (2), the sum of the weights across currencies for a given quarter be equal to one.
2. Compute the annual counterpart of the quarterly pattern series as

$$P_t^1 = \frac{\sum_{s=1}^4 P_{ts}^4}{4}, t = 1971...2005.$$

3. Compute the ratio of the annual target series to the newly created annual pattern series:

$$\rho_t^1 = \frac{T_t^1}{P_t^1}, t = 1971...2005.$$

4. Extend  $\rho_t^1$  backwards through 1970 with its 1971 value; extend  $\rho_t^1$  forward through 2006 with its 2005 value:  $\rho_{1970}^1 = \rho_{1971}^1$  and  $\rho_{2006}^1 = \rho_{2005}^1$ .
5. Apply a cubic spline to  $\rho_t^1$  using  $\rho_{1970}^1$  and  $\rho_{2006}^1$  as terminal conditions. This step yields a quarterly ratio series,  $\hat{\rho}_{ts}^4$ ,  $t = 1970...2006$ ,  $s = 1...4$ .
6. Obtain a first-round estimate of the quarterly target series as

$$\hat{T}_{ts}^4 = \hat{\rho}_{ts}^4 \cdot P_{ts}^4, t = 1970...2006, s = 1...4.$$

7. Compute the annual value implied by  $\hat{T}_{ts}^4$  as

$$\hat{T}_t^1 = \frac{\sum_{s=1}^4 \hat{T}_{ts}^4}{4}, t = 1970...2006.$$

8. Calculate the error between  $T_t^1$  and  $\hat{T}_t^1$ :  $\hat{\varepsilon}_t^1 = T_t^1 - \hat{T}_t^1$ ,  $t = 1970...2006$  with  $\hat{\varepsilon}_{1970}^1 = \hat{\varepsilon}_{2006}^1 = 0$ .
9. Construct a quarterly series of errors,  $\hat{\varepsilon}_{ts}^4$ , where the value of the error in each quarter of a

given year  $t$  is equal to  $\widehat{\varepsilon}_t^1$ . Thus  $\widehat{\varepsilon}_{ts}^4$  has the same value in all the quarters of a given year:

$$\begin{pmatrix} \widehat{\varepsilon}_{t,1}^4 \\ \widehat{\varepsilon}_{t,2}^4 \\ \widehat{\varepsilon}_{t,3}^4 \\ \widehat{\varepsilon}_{t,4}^4 \end{pmatrix} = \begin{pmatrix} \widehat{\varepsilon}_t^1 \\ \widehat{\varepsilon}_t^1 \\ \widehat{\varepsilon}_t^1 \\ \widehat{\varepsilon}_t^1 \end{pmatrix}, \quad t = 1970 \dots 2006.$$

10. Add  $\widehat{\varepsilon}_{ts}^4$  to  $\widehat{T}_{ts}^4$  to obtain a second-round estimate of the quarterly target series:

$$\widehat{T}_{ts}^4 = \widehat{T}_{ts}^4 + \widehat{\varepsilon}_{ts}^4, \quad t = 1970 \dots 2006, s = 1 \dots 4.$$

The last two steps ensure that the second-round estimate ( $\widehat{T}_{ts}^4$ ), when converted into an annual series, has the same values as our original target series ( $T_t^1$ ). Specifically,

$$\frac{\sum_{s=1}^4 \widehat{T}_{ts}^4}{4} = \frac{\sum_{s=1}^4 \widehat{T}_{ts}^4}{4} + \frac{\sum_{s=1}^4 \widehat{\varepsilon}_{ts}^4}{4} = \widehat{T}_t^1 + \widehat{\varepsilon}_t^1 = T_t^1.$$

Figure 16 shows that the profiles for the annual and quarterly values for WARP are identical.

## A.2 Chained and Geometric Aggregates: A Numerical Example

To illustrate the properties of these aggregates, we use a hypothetical numerical example in which there are three countries: the United States and two foreign countries: A high-price country,  $H$ , and a low-price country,  $L$ , and their bilateral relative prices are denoted as  $q_t^H$  and  $q_t^L$ . With this information, we compute the chained and geometric aggregates as

$$\begin{aligned} \frac{Q_t^c}{Q_{t-1}^c} &= \left( \frac{q_t^H}{q_{t-1}^H} \right)^{\omega_{Ht}} \cdot \left( \frac{q_t^L}{q_{t-1}^L} \right)^{\omega_{Lt}} \\ Q_t^g &= (q_t^H)^{\omega_{Ht}} \cdot (q_t^L)^{\omega_{Lt}}. \end{aligned}$$

The initial level of the chained aggregate is arbitrary and thus, to ease the comparison between the aggregates, we set the first period level of the chained equal to the first period level of the Geometric—that is,  $Q_{t=1}^g = Q_{t=1}^c$ . The values for the weights are those of the Broad Real index where we group industrial countries as the High-price country and emerging economies as the Low-price country. With this classification, the top panel of figure 17 shows the evolution of the weights. Notice that the data for the weight of the low-price country shows a relatively low starting value that has grown

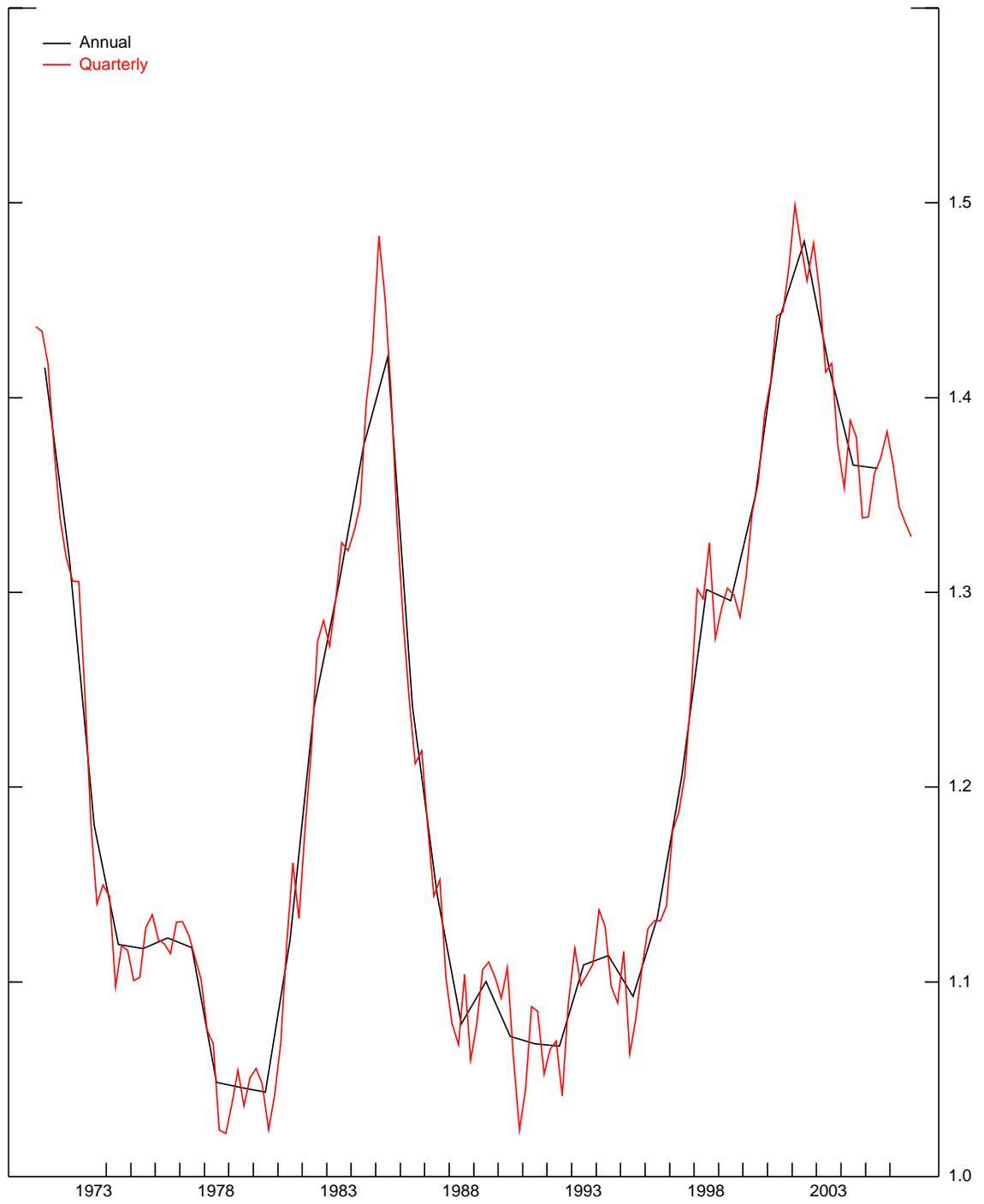


Figure 16: Geometric Aggregate of Bilateral PWT Relative Prices: Annual versus Quarterly

significantly over time.

To focus on the role of the interaction between changing weights and differentials in relative prices, we assume both nominal exchange rates against the dollar are set to one. With these assumptions, we consider two cases: fixed and changing bilateral relative prices.

**Case 1: Fixed Bilateral Relative Prices** We start out in period one with prices in country  $H$ , expressed in dollars, being 10 percent above those in the United States. Thus the bilateral relative price against  $H$ , denoted by  $q_t^H$  is 1 over 1.1 or 0.91:

$$q_t^H = \frac{P_t^u}{P_t^H} \cdot E_{\frac{u}{\$}} t = \frac{1}{1.1} \cdot 1 = 0.91.$$

For prices in country  $L$ , we assume they start out at one half those in the United States, so the dollar's bilateral relative price against  $L$  is 1 over 0.5 or 2:

$$q_t^L = \frac{P_t^u}{P_t^L} \cdot E_{\frac{u}{\$}} t = \frac{1}{0.5} \cdot 1 = 2.$$

The middle panel of figure 17 shows what happens if the individual bilateral relative prices remain fixed, but we let the weight of country  $L$  rise. Because bilateral relative prices are fixed, the chained index (shown in black) remains flat at its initial value. However, the geometric aggregate (in red) rises because the weight applied to the low-price country,  $q_t^L$ , is rising over time.

**Case 2: Changing Bilateral Relative Prices** Given the many ways in which one can assume hypothetical changes in relative prices, we impose several assumptions that are neutral with respect to aggregation. First, we assume that the fluctuations, in percent, of  $q_t^H$  and  $q_t^L$  are the same and that each bilateral relative price ends at its starting level. For the pattern of fluctuations, we use the ones registered by the Federal Reserve Board's broad real exchange rate. The bottom panel shows what happens if we let the bilateral relative prices move along with the weights. Both the chained and the geometric track the general movements in the bilateral relative prices; and in this rather special case, the chained aggregate returns to its initial value. The geometric, however, has an upward trend reflecting the increasing weight of the low price country. This is the main idea behind the geometric aggregate.

The main lesson from these graphs is that if we want to construct a measure of aggregate relative prices that reflects the evolving importance of countries with differing prices, then we need an aggregator that is sensitive to these differences.

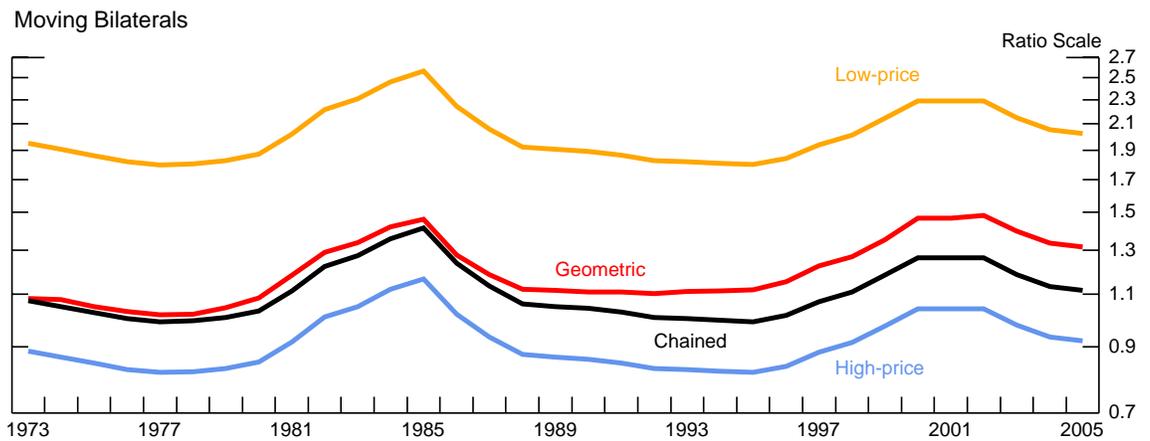
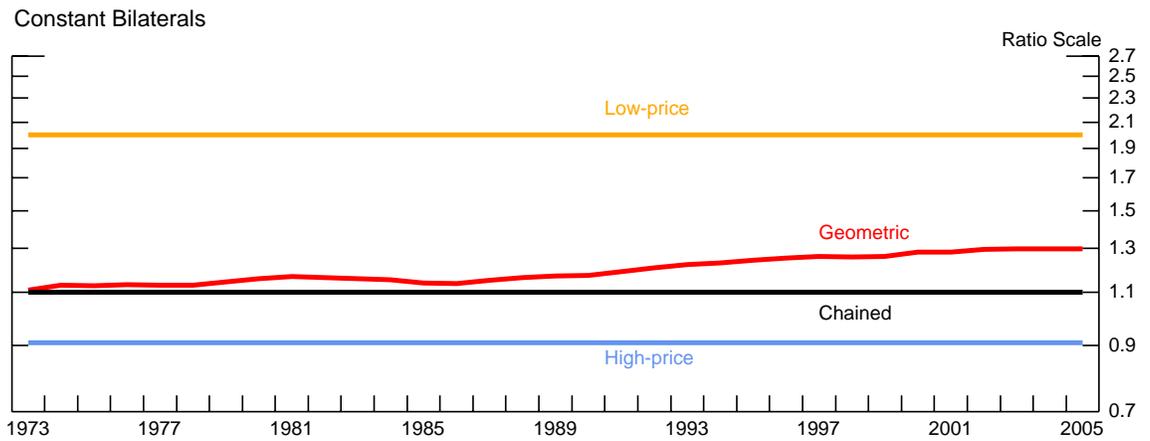
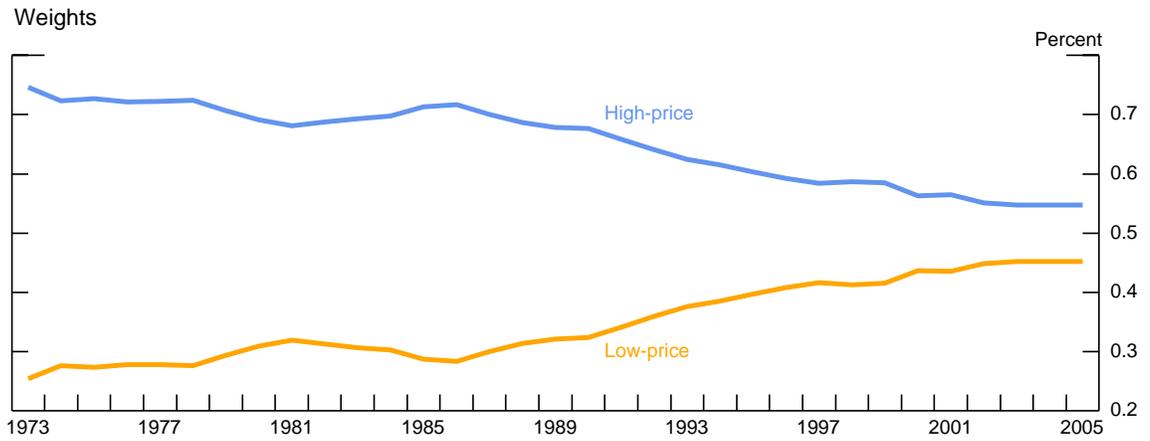


Figure 17: Hypothetical Measures of U.S. International Relative Price

### A.3 WARP and Substitutability Among Foreign Products

As noted in the text, the geometric aggregate  $Q^g$  is a particular case of

$$Q_t^{ces} = \left[ \sum_{i=1}^n \omega_{it} \cdot (q_{it})^{\sigma-1} \right]^{\frac{1}{\sigma-1}},$$

with  $Q^g = \lim_{\sigma \rightarrow 1} Q_t^{ces}$ . Evaluating the sensitivity of  $Q_t^{ces}$  to alternative values of  $\sigma$  is simple but interpreting the results requires an economic interpretation of  $\sigma$ . To that end, we rely on Varian (1984, p. 33) and assume that the bundle of foreign products,  $F$ , can be expressed as

$$F = \left[ \sum_{i=1}^n (a_i \cdot F_i)^{-\left(\frac{1-\sigma}{\sigma}\right)} \right]^{-1/\left(\frac{1-\sigma}{\sigma}\right)},$$

where  $F_i$  represents purchases of products from the  $i$ th country;  $a_i^{-\left(\frac{1-\sigma}{\sigma}\right)}$  is the distribution parameter that translates units of  $F_i$  into units of  $F$ ; and  $\sigma$  is the elasticity of substitution among foreign products: a large value of  $\sigma$  means that foreign products are highly substitutable whereas a small value of  $\sigma$  means the opposite.

To get from an assumption about the aggregate of purchases of foreign products to a measure of the U.S. international relative price, we follow Varian (1984, p. 33) and assume that  $F_i$  is determined so as to minimize the cost of attaining a given level of  $F$ ; the resulting price of the cost-minimizing bundle of foreign products,  $P_f$ , is

$$P_f = \left[ \sum_{i=1}^n \left( \frac{P_{fi}}{a_i} \right)^{1-\sigma} \right]^{\frac{1}{1-\sigma}}, \quad (14)$$

where  $P_{fi}$  is the (dollar) price of products from the  $i$ th country. Note that, in Varian's derivation, the distribution parameter,  $\left(\frac{1}{a_i}\right)^{1-\sigma}$ , depends on  $\sigma$ . Thus the familiar application of L'Hôpital's rule to equation (14) will not yield  $P_f$  as a geometric aggregate of purchases of foreign products as  $\sigma \rightarrow 1$ . To get that convergence one needs to assume that  $a_i = \omega_i^{1/(\sigma-1)}$ , where  $\omega_i$  is the  $i$ th country's weight in the Federal Reserve's Broad measure of the dollar. Thus, with this assumption, the resulting aggregate of foreign prices purchased by U.S. residents is given by

$$P_f = \left[ \sum_{i=1}^n \omega_i \cdot (P_{fi})^{1-\sigma} \right]^{\frac{1}{1-\sigma}}. \quad (15)$$

However, as written, equation (15) is not suitable for numerical analysis because data for  $P_{fi}$  are not available. To bypass this limitation, we develop an equivalent expression in terms of  $q_i = \frac{P_u}{P_{fi}}$  for

which data are available from the Penn World Tables. Thus, equation (15) can be re-expressed as

$$P_f = \left[ \sum_{i=1}^n \omega_i \cdot \left( P_u \cdot \frac{P_{fi}}{P_u} \right)^{1-\sigma} \right]^{\frac{1}{1-\sigma}} = \left[ \sum_{i=1}^n \omega_i \cdot (P_u)^{1-\sigma} \cdot (q_i)^{\sigma-1} \right]^{\frac{1}{1-\sigma}} = P_u \cdot \left[ \sum_{i=1}^n \omega_i \cdot (q_i)^{\sigma-1} \right]^{\frac{1}{1-\sigma}}.$$

The resulting international relative price of U.S. products is

$$\frac{P_u}{P_f} = \frac{P_u}{P_u \cdot \left[ \sum_{i=1}^n \omega_i \cdot (q_i)^{\sigma-1} \right]^{\frac{1}{\sigma-1}}} = \left[ \sum_{i=1}^n \omega_i \cdot (q_i)^{\sigma-1} \right]^{\frac{1}{\sigma-1}} \equiv Q^{ces}. \quad (16)$$

## A.4 Theoretical Measures of Competitiveness

This lengthy appendix constructs analytical analogues to the measures of competitiveness offered in the literature. Our goal is to show that they are just as subject to the influence of the prices of non-tradeables as WARP.

### A.4.1 Structure of Production and Trade

We assume that the world economy is divided into three country blocs: A foreign country of interest (country 1), the United States, and the Rest of the World (ROW). There are five goods, denoted as  $\Upsilon_i$ , that can be used either as intermediate inputs or as a final consumption: goods  $\Upsilon_1$  and  $\Upsilon_2$  are non-traded whereas goods  $\Upsilon_3$ ,  $\Upsilon_4$ ,  $\Upsilon_5$  are traded. Country 1 and the United States produce  $\Upsilon_1$ ,  $\Upsilon_2$ ,  $\Upsilon_3$ , and  $\Upsilon_4$ ; production is undertaken with labor and intermediates. The endowment of labor  $L$  is given and intermediates can be locally produced or imported. Product  $\Upsilon_5$  is not produced by these two countries because the marginal cost of production is assumed to exceed the world price. For the pattern of trade, we assume that country 1 exports  $\Upsilon_3$ , and imports  $\Upsilon_4$ ,  $\Upsilon_5$ ; that the United States exports  $\Upsilon_4$ , and imports  $\Upsilon_3$ ,  $\Upsilon_5$ ; and that ROW exports  $\Upsilon_5$  and imports  $\Upsilon_3$  and  $\Upsilon_4$ .

We denote  $\Upsilon_{ij}$  as the total supply of the product delivered from the  $i$ th industry to the  $j$ th productive sector. This supply can be made up of domestic production or imports— that is,  $\Upsilon_{ij} = \Pi_{ij} + m_{ij}$ , where  $\Pi_{ij}$  represents domestic production of the  $i$ th product purchased by the  $j$ th productive sector and  $m_{ij}$  represents imports of the  $i$ th product purchased by the  $j$ th productive sector.

We denote  $C_i$  as purchases of the  $i$ th product for final consumption. These purchases might be met by domestic production or by imports – that is,  $C_i = \Pi_{ic} + m_{ic}$ , where  $\Pi_{ic}$  represents purchases of domestic production of the  $i$ th product for final consumption and  $m_{ic}$  represents imports of the  $i$ th product for final consumption. Note that  $\Upsilon_{ij}$  and  $C_i$  treat domestic production and imports of the  $i$ th product as perfect substitutes for each other. Thus, we expect the law of one price to hold.

Finally, denoting  $X_i$  as exports of the  $i$ th product, the tableau of economic transactions is

|                   |                           | ← Users →                 |                 |                 |                 |                 |       |       |          |
|-------------------|---------------------------|---------------------------|-----------------|-----------------|-----------------|-----------------|-------|-------|----------|
|                   |                           | Productive Sector ( $j$ ) |                 |                 |                 | Final Demand    |       |       |          |
|                   |                           | $\Pi_1$                   | $\Pi_2$         | $\Pi_3$         | $\Pi_4$         | $C$             | $X$   | $M$   |          |
|                   | Labor                     | $L$                       | $L_1$           | $L_2$           | $L_3$           | $L_4$           |       |       |          |
| ↑                 | Industry 1                | $\Upsilon_1$              | $\Upsilon_{11}$ | $\Upsilon_{12}$ | $\Upsilon_{13}$ | $\Upsilon_{14}$ | $C_1$ | 0     | 0        |
| Input             | Industry 2                | $\Upsilon_2$              | $\Upsilon_{21}$ | $\Upsilon_{22}$ | $\Upsilon_{23}$ | $\Upsilon_{24}$ | $C_2$ | 0     | 0        |
| Suppliers ( $i$ ) | Industry 3                | $\Upsilon_3$              | $\Upsilon_{31}$ | $\Upsilon_{32}$ | $\Upsilon_{33}$ | $\Upsilon_{34}$ | $C_3$ | $X_3$ | 0        |
| ↓                 | Industry 4                | $\Upsilon_4$              | $\Upsilon_{41}$ | $\Upsilon_{42}$ | $\Upsilon_{43}$ | $\Upsilon_{44}$ | $C_4$ | 0     | $m_{4c}$ |
|                   | Non-Substitutable Imports | $\Upsilon_5$              | $\Upsilon_{51}$ | $\Upsilon_{52}$ | $\Upsilon_{53}$ | $\Upsilon_{54}$ | $C_5$ | 0     | $m_{5c}$ |

Entries along the first row denote deliveries of labor services to the  $j$ th productive sector. Deliveries by industries 1 and 2 are used as intermediate inputs and as final consumption. Deliveries by industry 3 are used as an intermediate input, as final consumption, and as exports; by assumption, good 3 is not imported. Deliveries by industry 4 are used as an intermediate input and as final consumption; these deliveries are augmented by imports of good 4. Deliveries of non-substitutable imports, good 5, are used as an intermediate input in all industries and in final consumption.

Entries along a column for the  $j$ th productive sector ( $j = 1, 2, 3, 4$ ) represent purchases of this sector of labor services and of products from the  $i$ th supplier of inputs ( $i = 1, 2, 3, 4, 5$ ). Entries in column  $C$  and  $X$  represents purchases for final consumption and for exports. Column  $M$  is not included in conventional presentations but we use it here to record imports for final consumption.

We use this tableau to establish the identity between uses and sources of the  $i$ th product. Specifically, the gross supply of the  $i$ th product facing substitutable imports,  $\Upsilon_i$ , is

$$\begin{aligned}
\Upsilon_i &= \underbrace{\overbrace{\sum_{j=1}^4 \Upsilon_{ij}}^{\text{intermediate}} + \overbrace{C_i}^{\text{consumption}} + \overbrace{X_i}^{\text{exports}}}_{\text{uses of the } i\text{th product}} = \sum_{j=1}^4 \underbrace{(\Pi_{ij} + m_{ij})}_{\Upsilon_{ij}} + \underbrace{(\Pi_{ic} + m_{ic})}_{C_i} + X_i \\
&= \sum_{j=1}^4 \Pi_{ij} + \Pi_{ic} + X_i + \sum_{j=1}^4 m_{ij} + m_{ic} = \underbrace{\overbrace{\sum_{j=1}^4 \Pi_{ij}}^{\text{domestic gross output}} + \overbrace{\sum_{j=1}^4 m_{ij} + m_{ic}}^{\text{imports}}}_{\text{sources of the } i\text{th product}}, \quad i = 1, 2, 3, 4.
\end{aligned}$$

Thus, domestic gross output is

$$\Pi_i = \sum_{j=1}^4 \Pi_{ij} + \Pi_{ic} + X_i. \tag{17}$$

The gross supply of non-substitutable imports ( $i = 5$ ) is

$$\Upsilon_5 = \sum_{j=1}^4 \underbrace{\Pi_{5j}}_{\text{zero}} + \underbrace{\Pi_{5c}}_{\text{zero}} + \underbrace{X_5}_{\text{zero}} + \sum_{j=1}^4 m_{5j} + m_{5c} = \sum_{j=1}^4 m_{5j} + m_{5c}.$$

**Technology** We assume a Leontieff production function for the  $j$ th productive sector ( $j = 1, 2, 3, 4$ ). Specifically, the labor requirement for producing one unit of the  $j$ th product is  $a_{0j} = \frac{L_j}{\Pi_j}$ , where  $\Pi_j$  is determined by equation (17). The direct requirement of the  $i$ th intermediate product needed for producing one unit of the  $j$ th product is  $a_{ij} = \frac{\Upsilon_{ij}}{\Pi_j}$ . The tableau in terms of technological coefficients is

|                           |              | Productive Sector ( $j = 1, 2, 3, 4$ ) |                      |                      |                      | Final Demand |       |          |
|---------------------------|--------------|--|----------------------|----------------------|----------------------|--------------|-------|----------|
|                           |              | $\Pi_1$                                | $\Pi_2$              | $\Pi_3$              | $\Pi_4$              | $C$          | $X$   | $M$      |
| Labor                     | $L$          | $a_{01} \cdot \Pi_1$                   | $a_{02} \cdot \Pi_2$ | $a_{03} \cdot \Pi_3$ | $a_{04} \cdot \Pi_4$ |              |       |          |
| Industry 1                | $\Upsilon_1$ | $a_{11} \cdot \Pi_1$                   | $a_{12} \cdot \Pi_2$ | $a_{13} \cdot \Pi_3$ | $a_{14} \cdot \Pi_4$ | $C_1$        | 0     | 0        |
| Industry 2                | $\Upsilon_2$ | $a_{21} \cdot \Pi_1$                   | $a_{22} \cdot \Pi_2$ | $a_{23} \cdot \Pi_3$ | $a_{24} \cdot \Pi_4$ | $C_2$        | 0     | 0        |
| Industry 3                | $\Upsilon_3$ | $a_{31} \cdot \Pi_1$                   | $a_{32} \cdot \Pi_2$ | $a_{33} \cdot \Pi_3$ | $a_{34} \cdot \Pi_4$ | $C_3$        | $X_3$ | 0        |
| Industry 4                | $\Upsilon_4$ | $a_{41} \cdot \Pi_1$                   | $a_{42} \cdot \Pi_2$ | $a_{43} \cdot \Pi_3$ | $a_{44} \cdot \Pi_4$ | $C_4$        | 0     | $m_{4c}$ |
| Non-substitutable Imports | $\Upsilon_5$ | $a_{51} \cdot \Pi_1$                   | $a_{52} \cdot \Pi_2$ | $a_{53} \cdot \Pi_3$ | $a_{54} \cdot \Pi_4$ | $C_5$        | 0     | $m_{5c}$ |

**Marginal Costs** Because we assume constant returns to scale, we express marginal costs in terms of prices and technological requirements. For example, the marginal cost in productive sector 3 is

$$c_3 = p_0 \cdot a_{03} + p_1 \cdot a_{13} + p_2 \cdot a_{23} + c_3 \cdot a_{33} + \bar{p}_4 \cdot a_{43} + \bar{p}_5 \cdot a_{53},$$

where  $p_0$  is the wage rate,  $p_i$  ( $i > 0$ ) is the price of the  $j$ th product, and a bar denotes the world price. Note that the industry's use of its own output is valued not at the market price but at the industry's own marginal cost. We now solve for marginal costs in terms of market prices and technology:

$$c_1 = (1 - a_{11})^{-1} \cdot [p_0 \cdot a_{01} + p_1 \cdot 0 + p_2 \cdot a_{21} + \bar{p}_3 \cdot a_{31} + \bar{p}_4 \cdot a_{41} + \bar{p}_5 \cdot a_{51}] \quad (18)$$

$$c_2 = (1 - a_{22})^{-1} \cdot [p_0 \cdot a_{02} + p_1 \cdot a_{12} + p_2 \cdot 0 + \bar{p}_3 \cdot a_{32} + \bar{p}_4 \cdot a_{42} + \bar{p}_5 \cdot a_{52}] \quad (19)$$

$$c_3 = (1 - a_{33})^{-1} \cdot [p_0 \cdot a_{03} + p_1 \cdot a_{13} + p_2 \cdot a_{23} + \bar{p}_3 \cdot 0 + \bar{p}_4 \cdot a_{43} + \bar{p}_5 \cdot a_{53}] \quad (20)$$

$$c_4 = (1 - a_{44})^{-1} \cdot [p_0 \cdot a_{04} + p_1 \cdot a_{14} + p_2 \cdot a_{24} + \bar{p}_3 \cdot a_{34} + \bar{p}_4 \cdot 0 + \bar{p}_5 \cdot a_{54}] \quad (21)$$

The generic expression for the marginal cost of the  $j$ th industry ( $j = 1, 2, 3, 4$ ) is

$$c_j = \underbrace{\sum_{i=0}^2 p_i \cdot f_{ij}}_{\text{non-tradeables}} + \underbrace{\sum_{i=3}^5 p_i \cdot f_{ij}}_{\text{tradeables}}, \quad (22)$$

where  $f_{ij} = \frac{a_{ij}}{1-a_{jj}}$  for  $i \neq j$  is the *total* amount of the  $i$ th product required to produce one unit of the  $j$ th product. The difference between total and direct requirements is the allowance of the  $j$ th product to serve as an input in the production of the  $j$ th product. Using equations (18)-(21), we compute the matrix of responses of marginal costs to changes in technological coefficients:

$$\left[ \frac{\partial c_j}{\partial a_{ij}} \right] = \begin{bmatrix} \frac{\partial c_1}{\partial a_{01}} & \frac{\partial c_2}{\partial a_{02}} & \frac{\partial c_3}{\partial a_{03}} & \frac{\partial c_4}{\partial a_{04}} \\ \frac{\partial c_1}{\partial a_{11}} & \frac{\partial c_2}{\partial a_{12}} & \frac{\partial c_3}{\partial a_{13}} & \frac{\partial c_4}{\partial a_{14}} \\ \frac{\partial c_1}{\partial a_{21}} & \frac{\partial c_2}{\partial a_{22}} & \frac{\partial c_3}{\partial a_{23}} & \frac{\partial c_4}{\partial a_{24}} \\ \frac{\partial c_1}{\partial a_{31}} & \frac{\partial c_2}{\partial a_{32}} & \frac{\partial c_3}{\partial a_{33}} & \frac{\partial c_4}{\partial a_{34}} \\ \frac{\partial c_1}{\partial a_{41}} & \frac{\partial c_2}{\partial a_{42}} & \frac{\partial c_3}{\partial a_{43}} & \frac{\partial c_4}{\partial a_{44}} \\ \frac{\partial c_1}{\partial a_{51}} & \frac{\partial c_2}{\partial a_{52}} & \frac{\partial c_3}{\partial a_{53}} & \frac{\partial c_4}{\partial a_{54}} \end{bmatrix} = \begin{bmatrix} \frac{p_0}{(1-a_{11})} & \frac{p_0}{(1-a_{22})} & \frac{p_0}{(1-a_{33})} & \frac{p_0}{(1-a_{44})} \\ \frac{c_1}{(1-a_{11})} & \frac{p_1}{(1-a_{22})} & \frac{p_1}{(1-a_{33})} & \frac{p_1}{(1-a_{44})} \\ \frac{p_2}{(1-a_{11})} & \frac{c_2}{(1-a_{22})} & \frac{p_2}{(1-a_{33})} & \frac{p_2}{(1-a_{44})} \\ \frac{\bar{p}_3}{(1-a_{11})} & \frac{\bar{p}_3}{(1-a_{22})} & \frac{c_3}{(1-a_{33})} & \frac{\bar{p}_3}{(1-a_{44})} \\ \frac{\bar{p}_4}{(1-a_{11})} & \frac{\bar{p}_4}{(1-a_{22})} & \frac{\bar{p}_4}{(1-a_{33})} & \frac{c_4}{(1-a_{44})} \\ \frac{\bar{p}_5}{(1-a_{11})} & \frac{\bar{p}_5}{(1-a_{22})} & \frac{\bar{p}_5}{(1-a_{33})} & \frac{\bar{p}_5}{(1-a_{44})} \end{bmatrix}. \quad (23)$$

**Producer's Markup** The producer markup of the  $j$ th productive sector is defined as

$$\mu_j = \begin{cases} \frac{p_j}{c_j} & \text{for } j = 1, 2 \\ \frac{\bar{p}_j}{c_j} & \text{for } j = 3, 4 \end{cases}. \quad (24)$$

**Sectoral Final Demand** Final demand for the  $i$ th product,  $Y_i$ , is derived as

$$\begin{aligned} Y_i &= \overbrace{\left[ \sum_{j=1}^4 a_{ij} \cdot \Pi_j + C_i + X_i \right]}^{r_i} - \left[ \sum_{j=1}^4 a_{ij} \cdot \Pi_j + m_{ic} \right] \\ &= [C_i + X_i] - m_{ic} \\ &= [\Pi_{ic} + m_{ic} + X_i] - m_{ic} = \Pi_{ic} + X_i, \quad i = 1, 2, 3, 4, \end{aligned} \quad (25)$$

**Nominal GDP** GDP is the *value* of final demands valued at prices that exclude indirect taxes:

$$Y = \sum_{i=1}^4 p_i \cdot Y_i. \quad (26)$$

#### A.4.2 Measures of Competitiveness

The structure developed above allows us to examine formally how the measures of competitiveness offered in the literature respond to technology shocks. To this end, we denote U.S. variables with a superscript  $u$ , and Country 1 variables with a superscript 1.

**Ratio of Marginal Costs of tradeable Products** The ratio of marginal costs of tradeable products is

$$\mathbb{C} = \frac{\sum_{j=3}^4 c_j^u \cdot Y_j^1}{\sum_{j=3}^4 c_j^1 \cdot Y_j^1}. \quad (27)$$

The denominator of  $\mathbb{C}$  is the value of tradeables in country 1 valued using that country's marginal costs; the numerator of  $\mathbb{C}$  is the value of the same bundle valued with U.S. marginal cost. Thus if  $\mathbb{C} > 1$ , then "Country 1 is said to be more competitive than the United States." The effect on  $\mathbb{C}$  of a decrease in U.S. productivity of non-tradeable product 1 ( $da_{1j}^u > 0$ ) is

$$d\mathbb{C} = \frac{1}{\sum_{j=3}^4 c_j^1 \cdot Y_j^1} \cdot \left[ \sum_{j=3}^4 \frac{\partial c_j^u}{\partial a_{1j}^u} \cdot Y_j^1 \cdot da_{1j}^u \right]. \quad (28)$$

Using the partial derivatives shown in equation (23),  $d\mathbb{C}$  can be re-expressed as

$$d\mathbb{C} = \frac{p_1^u}{\sum_{j=3}^4 c_j^1 \cdot Y_j^1} \cdot \left[ \frac{Y_3^1}{(1 - a_{33}^u)} \cdot da_{13}^u + \frac{Y_4^1}{(1 - a_{44}^u)} \cdot da_{14}^u \right] > 0.$$

This expression indicates that a decline in U.S. productivity of non-tradeables raises U.S. marginal costs relative to marginal costs in country 1. Thus,  $\mathbb{C}$  is not invariant to developments in the non-tradeable sector.

**Sectoral Ratio of Producers' Markups of Tradeables** Corden's measure of competitiveness for the  $j$ th tradeable product is the ratio of producers' markups:

$$\rho_j = \frac{\mu_j^1}{\mu_j^u}, \quad j = 3, 4.$$

This equation can be expressed as the U.S. marginal cost for the  $j$ th tradeable product relative to that in country 1:

$$\rho_j = \frac{\mu_j^1}{\mu_j^u} = \left( \frac{\bar{p}_j}{c_j^1} \right) / \left( \frac{\bar{p}_j}{c_j^u} \right) = \frac{c_j^u}{c_j^1}. \quad (29)$$

The effect of a decrease in the productivity of U.S. non-tradeable good 1 ( $da_{1j}^u > 0$ ) on  $\rho_j$  is

$$d\rho_j = \frac{1}{c_j^1} \cdot \frac{\partial c_j^u}{\partial a_{1j}^u} \cdot da_{1j}^u.$$

Using equation (23) yields

$$d\rho_j = \frac{1}{c_j^1} \cdot \frac{p_1^u}{(1 - a_{jj}^u)} \cdot a_{1j}^u \cdot \frac{da_{1j}^u}{a_{1j}^u}.$$

Recalling that  $f_{ij} = \frac{a_{ij}}{1 - a_{jj}}$  implies that

$$d\rho_j = \frac{1}{c_j^1} \cdot p_1^u \cdot f_{1j}^u \cdot \frac{da_{1j}^u}{a_{1j}^u} = \frac{c_j^u}{c_j^1} \cdot \frac{p_1^u \cdot f_{1j}^u}{c_j^u} \cdot \frac{da_{1j}^u}{a_{1j}^u} = \rho_j \cdot \nu_{j1}^u \cdot \frac{da_{1j}^u}{a_{1j}^u}, \quad (30)$$

where  $\nu_{j1}^u = \frac{p_1^u \cdot f_{1j}^u}{c_j^u}$  is the fraction of  $c_j^u$  accounted by cost of intermediate input 1. For the special case of  $\frac{da_{1j}^u}{a_{1j}^u} = 1$ , we get

$$\frac{d\rho_j}{\rho_j} = \nu_{j1}^u > 0, \quad (31)$$

This expression indicates that a decrease in U.S. productivity of non-tradeable product 1 raises  $\rho$ , even if  $\rho$  were an ideal measure of competitiveness for a given product.

## A.5 Empirical Implementation

**Computer-Automated Specification Algorithm** The computer-automated algorithm used here (*PcGets*), developed by Hendry and Krolzig (2001) and Krolzig and Hendry (2001), combines ordinary least squares with a selection strategy that is implemented in four stages:

1. Estimate the parameters of a general formulation—equation (13) for example—and test for congruency (e.g., white-noise residuals and parameter constancy).
2. Implement multiple “simplification paths” simultaneously. One simplification path could get started by excluding the least significant variable whereas another simplification path could get initiated by excluding a block of variables that are jointly insignificant.
3. Test whether the specification from a simplification path is congruent. If it is, then implement

another round of simplifications and re-test for congruency; continue this process until the specification violates congruency. In that case, the algorithm selects the immediately prior specification and labels it *Final model*.

4. Collect the Final models from all simplification paths and apply encompassing tests to them. The specification that encompasses all others becomes the *Specific model*. If there is no single encompassing model, then the algorithm forms a “union” model using the variables from all of the Final models and re-starts the specification search from step (2). If this strategy fails to yield a single Specific model, then the algorithm applies three information criteria (Akaike, Schwarz, and Hannan-Quinn) to the Final models and selects the one that minimizes all these criteria; that model becomes the Specific model. There is no guarantee that reliance on these three criteria will yield a unique model. In that event, the user specifies a criteria ranking to settle the conflict; this paper uses the Akaike Information Criterion. Otherwise, the algorithm fails to find a Specific model.

**Monte-Carlo Distributions of Long-run Elasticities** The distributions of  $\hat{\eta}_x^j = -\frac{\hat{\theta}_y^j}{\hat{\theta}_x^j}$  and  $\hat{\varepsilon}_x^j = -\frac{\hat{\theta}_p^j}{\hat{\theta}_x^j}$  are not known in advance because they are the ratios of normal variables which lack a well known distribution. Thus we generate the distributions of these elasticities in three steps:

1. Generate the  $k$ th drawing of  $\hat{\theta}^j$  ( $j = geo, c, cpi$ ) as

$$\hat{\theta}_k^j \equiv \hat{\theta}_{3 \times 1}^j + \hat{\Gamma}_{3 \times 3}^j \cdot \xi_k, \quad \hat{\Omega}^j = \hat{\Gamma}^j \cdot \hat{\Gamma}^j, \quad \xi_k \sim N(0, I_3), \quad k = 1, \dots, 1000,$$

where

$$\hat{\theta}^j \equiv \begin{pmatrix} \hat{\theta}_x^j \\ \hat{\theta}_y^j \\ \hat{\theta}_p^j \end{pmatrix}, \quad \hat{\Omega}^j = \begin{bmatrix} \hat{v}ar(\hat{\theta}_x^j) & \hat{c}ov(\hat{\theta}_x^j, \hat{\theta}_y^j) & \hat{c}ov(\hat{\theta}_x^j, \hat{\theta}_p^j) \\ \hat{c}ov(\hat{\theta}_x^j, \hat{\theta}_y^j) & \hat{v}ar(\hat{\theta}_y^j) & \hat{c}ov(\hat{\theta}_y^j, \hat{\theta}_p^j) \\ \hat{c}ov(\hat{\theta}_x^j, \hat{\theta}_p^j) & \hat{c}ov(\hat{\theta}_y^j, \hat{\theta}_p^j) & \hat{v}ar(\hat{\theta}_p^j) \end{bmatrix},$$

and  $\hat{\Gamma}^j$  is the lower-triangular Cholesky decomposition of  $\hat{\Omega}^j$ . We use the same seeds of the random-number generator across the various measures of relative export prices to ensure comparability of the results.

2. Compute the long-run elasticities associated with the  $k$ th drawing as

$$\hat{\eta}_{x,k}^j = -\frac{\hat{\theta}_{y,k}^j}{\hat{\theta}_{x,k}^j} \quad \text{and} \quad \hat{\varepsilon}_{x,k}^j = -\frac{\hat{\theta}_{p,k}^j}{\hat{\theta}_{x,k}^j}.$$

3. Use the 1000 drawings for each elasticity to generate the associated empirical densities.

Figures 18 and 19 show the densities for income and price elasticities associated with the Specific formulations of table 1. The results show the median and the bounds associated with 95% confidence interval; note the lack of symmetry in the underlying distributions.

### **Data for Modeling Aggregate U.S. Exports**

Data for U.S. exports of goods and services ( $X$ ), measured in chained 2000 dollar and on a NIPA-basis, come from Bureau of Economic Analysis, *Survey of Current Business*, table 4.2.6.

Data for the deflator of U.S. exports of goods and services ( $P_x$ ), come from Bureau of Economic Analysis, *Survey of Current Business*, table 4.2.4.

We measure economic activity outside the United States as a geometric weighted average of the real GDP indexes for all the countries included in the Federal Reserve's Broad measure of the effective value of the dollar—that is,  $Y_t^* = \prod_{i=1}^N (Y_{i,t})^{w_{it}}$ , where  $Y_{i,t}$  is the index of real GDP of the  $i$ th country;  $w_{it}$  is the bilateral trade weight that excludes the contribution of imports and includes the role of bilateral export and third-country markets. Data for each country's real GDP come from national sources and from the IMF.

Data for the weights come from the Federal Reserve's statistical release H10:

<http://www.federalreserve.gov/releases/h10/Weights/>

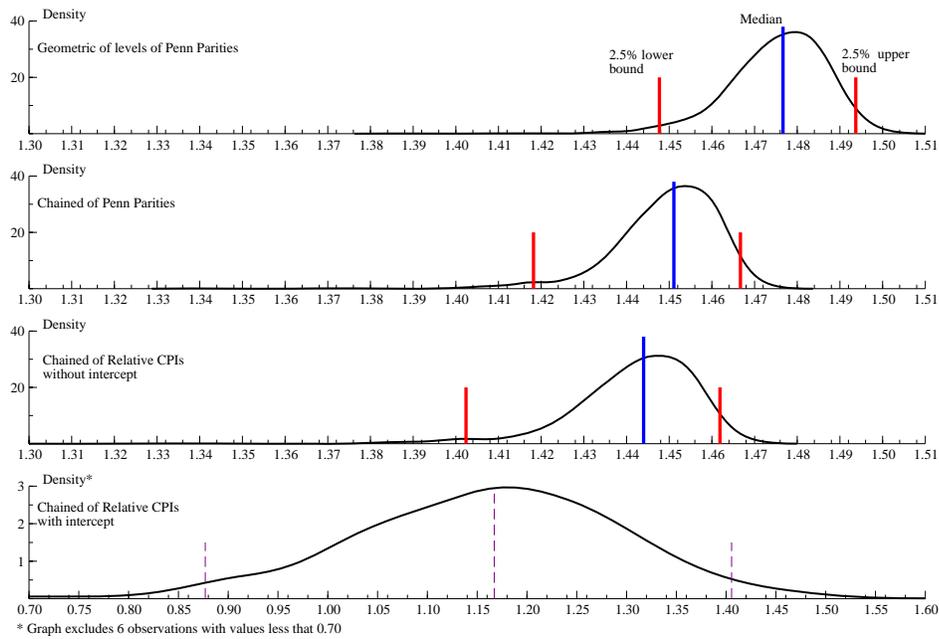


Figure 18: Monte Carlo Densities of Long-run Income Elasticity

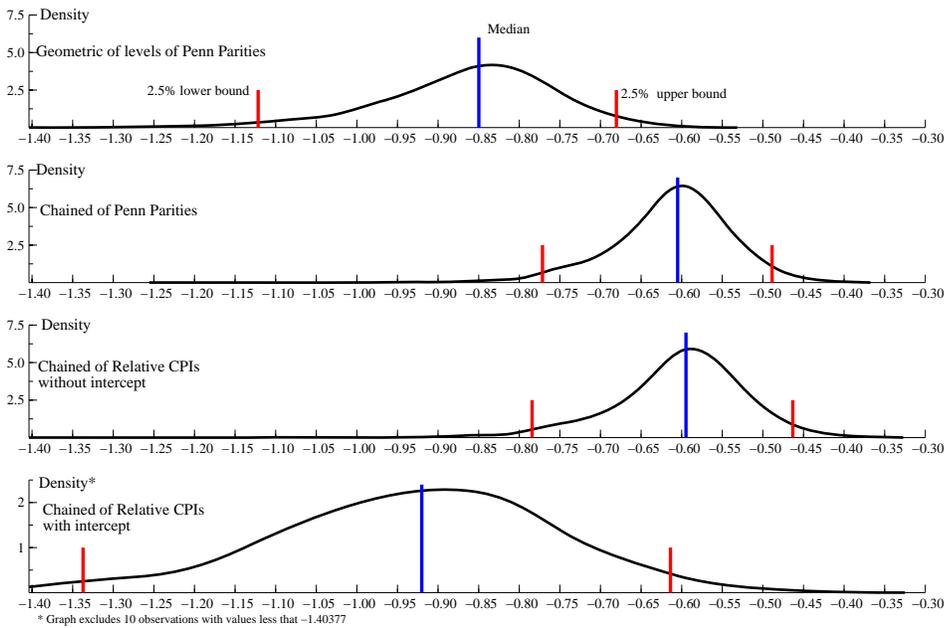


Figure 19: Monte Carlo Densities of Long-run Price Elasticity

Table A1: Augmented Dickey-Fuller Tests for Stationarity

|                       | D-lag | t-adf   | coefficient<br>on level of<br>Dep.<br>Variable | SER    | AIC    |
|-----------------------|-------|---------|--|--------|--------|
| ln Y*                 | 3     | -1.125  | 0.9987   | 0.0041 | -10.98 |
|                       | 2     | -1.126  | 0.9987   | 0.0040 | -10.99 |
|                       | 1     | -1.172  | 0.9987   | 0.0040 | -11.01 |
|                       | 0     | -2.048  | 0.9973   | 0.0048 |        |
| ln X                  | 3     | -0.7305 | 0.9974   | 0.0240 | -7.425 |
|                       | 2     | -0.7337 | 0.9974   | 0.0239 | -7.438 |
|                       | 1     | -0.7605 | 0.9972   | 0.0245 | -7.394 |
|                       | 0     | -0.9199 | 0.9966   | 0.0248 |        |
| ln rpx <sup>g</sup>   | 3     | -1.077  | 0.9861   | 0.0235 | -7.463 |
|                       | 2     | -0.925  | 0.9880   | 0.0238 | -7.448 |
|                       | 1     | -0.9347 | 0.9880   | 0.0237 | -7.463 |
|                       | 0     | -0.7859 | 0.9898   | 0.0240 |        |
| ln rpx <sup>c</sup>   | 3     | -0.2692 | 0.9976   | 0.0232 | -7.488 |
|                       | 2     | -0.1053 | 0.9991   | 0.0235 | -7.469 |
|                       | 1     | -0.1198 | 0.9989   | 0.0234 | -7.484 |
|                       | 0     | 0.03021 | 1.0003   | 0.0236 |        |
| ln rpx <sup>cpi</sup> | 3     | -0.3818 | 0.9968   | 0.0228 | -7.521 |
|                       | 2     | -0.2412 | 0.9980   | 0.0231 | -7.506 |
|                       | 1     | -0.2596 | 0.9978   | 0.0230 | -7.521 |
|                       | 0     | -0.1221 | 0.9990   | 0.0233 |        |

Sample is 1972Q2 to 2004Q4 (NOBS=131)

ADF regression includes a constant

The 5% rejection value is -2.88; the 1% rejection value is -3.48

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