Abstract

Estimates of the elasticity of substitution between domestic and foreign varieties are small in macroeconomic data, but substantially larger in disaggregated microeconomic studies. This may be an artifact of heterogeneity. We use disaggregated multilateral trade data to structurally identify imports elasticities in the US. We spell out a partial equilibrium model to aggregate them adequately at the country level. We compare aggregate elasticities that impose equality across sectors, to estimates allowing for heterogeneity. The former are similar in value to conventional macroeconomic estimates; but they are more than twice larger -up to 5 - with heterogeneity. The parameter is central to calibrated models in most of international economics. We discuss the difference our corrected estimate makes in various areas of international economics, including the dynamics of external balances, the international transmission of shocks, international portfolio choice and optimal monetary policy.

Keywords: Trade Elasticities, Aggregation, Calibration, Global Imbalances, International Transmission, International Portfolio, Monetary Policy.

JEL Classification: F41, F32, F21.
1 Introduction

The substitutability between domestic and foreign goods is central to most calibrated models in international economics. Depending on the value assigned to the parameter, the predictions of virtually any calibration exercise with an international dimension change quantitatively, sometimes dramatically. Naturally, the parameter value is key in models that seek to quantify the magnitude of a change in international prices consistent with a rebalancing of external balances, and the associated welfare consequences, as in Obstfeld and Rogoff (2005). It is also a crucial element to the mechanisms that underpin the international diffusion of shocks, as in Kose and Yi (2006) or Corsetti, Dedola and Leduc (2008); or indeed to the extent of international risk sharing in Cole and Obstfeld (1991), and the composition of international portfolio holdings, as in Coeurdacier (2005). Models of international price differences have predictions that rest on the parameter value as well, as demonstrated in Mirdigan (forthcoming) or Atkeson and Burstein (forthcoming). The same is true of the importance of exchange rates in the optimal conduct of monetary policy, as in Galí and Monacelli (2005).

This is quite simply one of the most important parameters in international economics. Unsurprisingly, its calibrated value draws from literally decades of empirical work. Unfortunately, little consensus has emerged from the effort, except for two broad conclusions. First, finely disaggregated good-level quantities are more responsive to (international) relative prices than aggregates. Second, there are enormous differences between goods. Long time ago, Orcutt (1950) referred to an “elasticity pessimism”, which he related to the gap between the low observed volatilities in aggregate quantities and the high volatility of international relative prices. He already conjectured that aggregates could obscure more responsive quantities at the microeconomic level. Here we ask whether this very heterogeneity may not actually be cause for optimism. We propose a correction of the elasticity of substitution between domestic and foreign macroeconomic quantities that accounts for heterogeneity at the microeconomic level.

Why should such a correction be meaningful? The robust finding that microeconomic studies uncover systematically higher estimates than aggregate data strongly points to the possibility of an aggregation bias. But this does not mean aggregate estimates are biased: after all it is the response of aggregate quantities that macroeconomists are interested in, and they
are undeniably much smoother than international prices. Rather, it means the assumption that all quantities are equally substitutable and adjust homogeneously to changes in relative prices is not supported by the data. As a result there might be systematic differences between the aggregate responsiveness of traded quantities and the preference parameter it is meant to capture. It is of course the latter that should enter calibrated models. The bias we discuss matters for calibration purposes.

Our first goal is to develop a general model telling us how to properly aggregate micro-economic estimates. We construct a measure of aggregate substitutability consistent with a representative agent choosing between aggregates of domestic and foreign quantities, and that can also accommodate the well documented fact that substitutability is vastly different across goods and sectors. We simulate the standard homogeneity assumption, under which elasticities are identical across sectors. This puts our theory to the test of its ability to reproduce conventional macroeconomic estimates under conventional macroeconomic assumptions. At the same time, the model maps out an index of substitutability between domestic and foreign aggregates, and a weighted average of its disaggregated counterparts. We let theory tell us which weights must be used for aggregation.

Our second goal is to estimate the disaggregated elasticities thus weighted. To do so, we borrow from a methodology introduced by Feenstra (1994a) and recently implemented by Broda and Weinstein (2006). Substitutability between two varieties is identified via the observed cross-country variation in the trade flows towards a given importer. We adapt the approach to our purposes imposing an Armington assumption. Demand at a sectoral level is given by an aggregator of domestic and foreign goods varieties. Crucially, the substitutability between two imported varieties is assumed to be identical to that between foreign and domestic types. Domestic and foreign varieties do differ in terms of transport costs and preferences, but not substitutability. It then becomes possible to use the properties of traded quantities and prices to identify the substitutability of interest. In comparison with conventional approaches, this provides estimates that are structural, and do not fall victim to the endogeneity concerns that plague any regression of (relative) quantities on (relative) prices. These are rather frequent in empirical work estimating trade elasticities.1

1See the seminal contributions of Orcutt (1950) or H.S. Houthakker and Stephen Magee (1969). Many
The Armington assumption requires that imports from different countries be imperfectly substitutable varieties. The hypothesis is increasingly palatable as the granularity of the data augments. We choose to use disaggregated, multilateral trade data from the Base Analytique du Commerce International (BACI), released by the Centre d’Etudes Prospectives et d’Informations Internationales (CEPII), and available at the HS6 level for a large cross-section of countries. Unlike Feenstra (1994a) or Broda and Weinstein (2006), we do not estimate elasticities at the most disaggregated level in the main body of the text, but rather partition our data into 56 industries where we implement our methodology. This is done for practical purposes and corresponds to the assumption that substitutability be identical between all the (HS6) varieties of a given (ISIC) industry. Presumably this assumes some heterogeneity away, and if we are right, creates a bias. We conjecture the cross section of elasticities is vastly more dispersed between ISIC industries, than between (HS6) goods belonging to the same sector. Our assumption is probably of second order relative to one - immediately implied by any macroeconomic estimate - imposing equal elasticities across all sectors in the economy.\(^2\)

In macroeconomic applications, calibration exercises typically favor values of the parameter that correspond to aggregate estimates. This is done for lack of a consensus, often because they are construed as “plausible mid-points” to the wide range of estimates the literature has uncovered. For instance, Obstfeld and Rogoff (2005) use a value of 2; Backus, Kehoe and Kydland (1994) use 1.5. The elasticity is (crucially) unitary in Cole and Obstfeld (1991), 0.9 in Heathcote and Perri (2008), 1.5 in Chari, Kehoe and McGrattan (2002), and set between 0.6 and 2 in Coeurdacier, Kollmann and Martin (2007). When all elasticities are forced to be equal across sectors, our approach generates aggregate estimates around 2.5 for the US. This is firmly within the ballpark of the calibrated values used in the macroeconomic literature, and strongly suggests there is nothing special to our data or our approach relative to existing empirical work. We seem to identify the object of interest to many international economists.

With heterogeneity however aggregate elasticity estimates more than double, with values in the US up to 5. This reflects the well known fact that microeconomic elasticities tend to be approaches have been developed since to address the endogeneity issue, but few are structural in nature.

\(^2\)In fact, we verify our main conclusions obtain with elasticities estimates at the HS6 level. But this requires assumptions on the weights used for aggregation, which are not directly observable at such a refined level.
high on average, and also quite dispersed. It is also the result of an appropriate, theory-implied weighting of disaggregated estimates. Does this correction matter economically? We discuss some illustrations in areas as diverse as the resolution of global imbalances, the international transmission of shocks, international risk sharing, portfolio choice, models of the real exchange rate and optimal monetary policy. In short, the parameter is central to most of international macroeconomics. Whether it is in fact 1 or 5 does make a quantitative difference, and calibrated equilibrium responses often change sizably. In some instances, the correction is also relevant qualitatively: the international response to productivity shocks, the justification for a home bias in portfolio investment or the Pareto-ranking of different macroeconomic policies can all alter profoundly with an elasticity substantially above one.

The impact of heterogeneity is significant and robust. Differences between standard and corrected elasticity estimates are statistically significant at conventional confidence levels, and they prevail for a variety of alternative measures or econometric procedures. Our conclusions withstand controls for common components in prices and quantities, and a battery of alternative data sources used to aggregate up microeconomics estimates. Point estimates are minimally affected, and heterogeneity always retains an economically and statistically significant impact. Our estimates of the aggregate substitutability between domestic and foreign goods in the US suggest it is closer to 5 than to unity.

In what follows, we first present the partial equilibrium model used to guide the aggregation of industry specific elasticities. Section 3 discusses the identification of sector specific parameters, their aggregation and the data involved. We also describe our adaptation of Feenstra’s approach. Section 4 reports our results, and document their relevance in recent standard models of international trade in goods and assets. Section 5 concludes.

2 Trade Elasticities: Practice ahead of Theory

We open with a summary of the empirical literature concerned with estimating trade elasticities. We discuss common practice, and how end estimates tend to be larger on average but heterogeneous in disaggregated data. We then lay out a partial equilibrium theory which we
use to map out conventional elasticity estimates with the utility parameter of interest in macroeconomic calibration exercises. The model tells us how to compute aggregate trade elasticities in ways that may or may not allow for heterogeneity at the good’s or sector’s level.

2.1 Practice

The substitutability between foreign and domestic goods can be directly inferred from estimates of the price elasticity of imports. As will become clear, theory implies a simple linear relation between the two. Estimating trade elasticities in general is an old business in economics. Early contributions used simple regression analysis to evaluate the response of aggregate imported quantities to changes in international relative prices. Houthakker and Magee (1969) for instance, estimate import elasticities between -0.5 and -1.5 depending on the economy considered. The decades that followed saw a flurry of econometric refinements, that purported to alleviate the endogeneity issues that go with having prices as independent variables. Thus, Marquez (1990) implements a frequency domain estimator, Gagnon (2003) instruments import prices using the real exchange rate, and Hooper, Johnson and Marquez (1998) use co-integration techniques. In most instances, estimates based on macroeconomic data were found to be weakly negative, not always significant, and certainly no larger than 2 in absolute value.

In contrast, the results obtained from disaggregated information are centered around higher averages, and substantially more heterogeneous. For instance, Blonigen and Wilson (1999) document elasticities between zero and 3.52 in 146 US sectors. Microeconomic studies also open the door to instrumentation strategies where changes in international relative prices can be ascribed to events, such as trade liberalizations, whose magnitude and timing are arguably exogenous to each market’s circumstances. Thus, using NAFTA and detailed information on good specific tariff changes, Head and Ries (2001) find disaggregated elasticities between 8 and 12, while Romalis (2007) documents estimates between 4 and 13.

A comparison of two recent papers crystallizes the importance of aggregation. Corsetti, Dedola and Leduc (2008) implement a General Method of Moments (GMM) on a macroeco-

\footnote{Most papers in this literature also take interest in the response of quantities traded to changes in income. We do not discuss this parameter here.}
nomic model to conclude the properties of aggregate data are best matched with an elasticity approximately equal to 0.5. Using a similar GMM approach, Bernard, Eaton, Jensen, and Kortum (2003) show their theory of plant-level behavior is best able to match observed data for values of the parameter around 4.

In short, there is every indication that trade elasticities are higher on average at a disaggregated level. This comes as no surprise. More than fifty years ago, Orcutt (1950) already remarked that “it is widely recognized that the demand schedule for the product of an individual producer has, in general, far greater price elasticity than the aggregate demand schedule for the entire output of the product” (page 118). The importance of aggregation in the context of trade elasticities has a straightforward, economic intuition.

Ruhl (2005) proposes an alternative. Cross-sectional elasticity estimates are naturally higher, for they embed firm dynamics, and the associated adjustments in the quantities produced. Time series data, in contrast, tend to focus on high or medium frequency developments, and may overlook most entry or exit decisions. But in practice, disaggregated datasets tend to be cross-sectional, whereas aggregate ones have information over time. The importance of aggregation may therefore be an artefact of data availability and differences in econometric methodologies. The parameters estimated in micro- and macro-economic studies are in fact fundamentally different, since in practice they do not capture the extensive margin to the same extent.

In what follows, we are careful to accommodate this possibility. Both our corrected and conventional elasticity estimates arise from the same dataset. Both are therefore affected by a putative extensive margin to exactly the same extent. In fact, our homogeneous and heterogeneous estimators are conceptually similar, and they use the same dimension in the same data. While each estimation may be affected by unaccounted firm dynamics, the discrepancy between the two cannot.\textsuperscript{4}

\textsuperscript{4}Feenstra (1994a) actually shows how the empirical approach accommodates time-varying number of firms in each exporting economy. We clarify the argument when we describe the estimation.
2.2 Theory

Consumption in the domestic economy is an aggregate of imperfectly substitutable goods \( k = 1, ..., K \). Utility is given by

\[
C = \left[ \sum_{k \in K} (\alpha_k \ C_k^{\gamma_k^{-1}}) \right]^{\gamma_k^{-1}}
\]

with \( \gamma \) the constant elasticity of substitution between goods, and \( \alpha_k \) an exogenous preference parameter. Consumption in each sector is derived from a continuum of varieties of good \( k \), that may be imported or not, as in

\[
C_k = \left[ \sum_{i \in I} (\beta_{ki} \ c_{ki}^{\sigma_k^{-1}} + (\beta_{kd} \ c_{kd}^{\sigma_k^{-1}})) \right]^{\sigma_k^{-1}}
\]

where \( i \) indexes imported types and \( d \) is the domestic variety of good \( k \). Crucially, the elasticity of substitution \( \sigma_k \) is specific to each industry, and assumed identical across all varieties, imported or not. \( \beta_{ki} \) lets preferences vary exogenously across varieties, reflecting for instance differences in quality. The sectors that verify \( \beta_{ki} = 0 \) for all \( i \neq d \) will effectively be non-traded.

This structure of demand is classic in international economics. The key assumption for our purposes is equal substitutability between two varieties, no matter their origin. Introducing the assumption is largely what opened the door to the New Trade literature, pioneered by Krugman (1980), and laid the foundation for the more recent models of trade with heterogeneous firms, starting with Melitz (2003).

The representative maximizing agent chooses her consumption allocation keeping in mind that all imported varieties incur a transport cost \( \tau_{ki}, i \in I \). Utility maximization implies that the quantities imported from country \( i \) in each sector \( k \) are given by

\[
c_{ki} = \beta_{ki}^{\sigma_k^{-1}} \left( \frac{P_i^*}{P_k} \right)^{-\sigma_k} \alpha_k^{\gamma_k-1} \left( \frac{P_k}{P} \right)^{-\gamma} C, \ i \neq d
\]

\(^5\)We could introduce without loss of generality an additional price wedge, reflecting distribution costs that presumably affect both domestic and foreign varieties. This would merely add some notation, but no further insight. In the empirics, the price of each variety is measured Free on Board, i.e. net of both retail and transportation costs.
with

\[ p_{ki}^* = \tau_{ki} E_i p_{ki} \]

\[ P_k = \left[ \sum_{i \in I} \left( \frac{p_{ki}^*}{\beta_{ki}} \right)^{1-\sigma_k} + \left( \frac{p_{kd}}{\beta_{kd}} \right)^{1-\sigma_k} \right]^{\frac{1}{1-\sigma_k}} \]

\[ P = \left[ \sum_{k \in K} \left( \frac{P_k}{\alpha_k} \right)^{1-\gamma} \right]^{\frac{1}{1-\gamma}} \]

Here, \( p_{ki} \) denotes the free on board price of variety \( i \) of good \( k \), labeled in country \( i \)'s currency, and \( E_i \) is the bilateral exchange rate with exporter \( i \). We assume producer currency pricing, and impose full pass through. In practice this focuses on estimates of the long run substitution elasticity, in that it excludes short-run dynamics in prices and quantities that stem from nominal rigidities.

2.3 Aggregation

We now ask our model how the estimated response of aggregate quantities to changes in aggregate international relative prices is affected depending on whether heterogeneity in \( \sigma_k \) is permitted. For this to be a meaningful experiment in a model with multilateral trade at the industry level, we consider disturbances to international relative prices of a specific kind. First, we focus on changes in all relative prices, across all sectors \( k \) in \( K \). This assumes away reallocation of demand across industries, with relative prices changing by the same amount in all sectors. It is relative quantities whose responses may be heterogeneous, which in turn may obscure aggregate estimates.

Second, we focus on uniform shocks to the international price of domestic goods, across all exporters \( i \) in \( I \). This assumes away reallocation of demand across source exporting economies, with relative prices changing identically in all markets. We do this for practical reasons, so that the multilateral dimension of the model collapses into a two-country version, and we can interpret our estimate as capturing the substitutability between composite goods in the domestic economy and in the rest of the world.\(^6\)

\(^6\)The second assumption is made for convenience. The intuition remains the same if we focus on a change in (all) relative prices between the domestic economy and a specific exporter \( i \). The data needed to perform
A natural candidate is a domestic shock to the nominal exchange rate driven, for instance, by a monetary disturbance. This changes the international price of domestic goods by an identical amount across all sectors $K$ and exporters $I$. Now, the conventional approach to identifying the substitutability $\eta$ between domestic and foreign varieties using macroeconomic data consists in estimating the response of imported quantities to changes in aggregate relative prices. In the model, this is equivalent to

$$\eta = \frac{\partial \left(\sum_k \sum_i p_{ki}^* c_{ki}\right)}{\partial E} \cdot \frac{E}{\sum_k \sum_i p_{ki}^* c_{ki}}$$

where $E$ denotes the effective exchange rate.\footnote{Given our assumption of a uniform shock across all $i$, this is equivalent to $\frac{\partial \left(\sum_k \sum_i p_{ki}^* c_{ki}\right)}{\partial E_i} \cdot \frac{E_i}{\sum_k \sum_i p_{ki}^* c_{ki}}$, since $\frac{\partial E_i}{\partial E} \cdot E_i = 1$.} We have taken into account that imported quantities (especially at the aggregate level) are expressed in nominal terms in the conventional datasets on the universe of international trade flows (ComTrade or BACI). We also let these nominal values be expressed in the end consumer’s currency, and inclusive of transport costs or tariffs (i.e. Cost, Insurance, Freight). This is innocuous: we later show that the estimation procedure is robust to the presence or otherwise of $\tau_{ki}$ in measured trade flows.

Equation (1) can now be used to obtain a model-implied expression for $\eta$. In particular,

$$\eta = 1 - \sum_{k \in K} n_k \sigma_k + \sum_{k \in K} n_k w_k^M (\sigma_k - \gamma) + \gamma \sum_{k \in K} n_k w_k^M \cdot w_k$$

(2)

with

$$n_k \equiv \frac{\sum_{i \in C_k} p_{ki}^* c_{ki}}{\sum_{k \in K} \sum_{i \in C_k} p_{ki}^* c_{ki}}$$

$$w_k^M \equiv \frac{\partial P_k E}{\partial E P_k} = \frac{\sum_{i \in I} p_{ki}^* c_{ki}}{P_k C_k}$$

$$w_k^M \cdot w_k \equiv \frac{\partial P E}{\partial E P} = w_k^M \cdot \frac{P_k C_k}{P C}.$$
The aggregate elasticity \( \eta \) contains three terms. First, a weighted average of industry-specific elasticities, with weights that correspond to the importance of sector \( k \) in overall imports. The second term reflects the response of the industry specific price index \( P_k \). Since by assumption the relative price of good \( k \) changes identically across all source economies \( i \), the composition of the ideal price index in sector \( k \) changes significantly in response to the shock considered. The third term reflects the adjustment of the aggregate price index \( P \), in response to an aggregate shock to relative prices affecting all sectors. In the words of Orcutt (1950), \( \eta \) denotes a total elasticity, which “indicates the percentage change in the quantity imported per percentage change in the price of the import, with allowance made for the full response of [...] competitors” (page 118).

The second and third terms in equation (2) exist because of our focus on an aggregate multilateral shock in relative prices. If instead the shock considered were microeconomic in nature and focused on a specific good \( k \)-a change in tariff- then under standard atomistic assumptions and large \( K \), the third term in equation (2) would disappear. Similarly, \( w_i^M \) is non zero because we assume all relative prices across all exporting economies change identically; for large values of \( I \), a change in a given bilateral exchange rate would have negligible effect on \( P_k \).

Two differences arise between the aggregate elasticity of imports to international prices and the preference parameter it is meant to estimate. One is related to the possibility of sectoral heterogeneity in the elasticity parameter, i.e. \( \sigma_k \neq \sigma_s, k \neq s \). The other has to do with the effect of macroeconomic shocks on (sectoral and aggregate) price indices. This possibility is well known in the literature. Orcutt (1950) introduces a partial elasticity measure, that assumes aggregate price responses away and captures “what would happen if the price of domestic substitutes were held constant” (page 118).

Focus for now on heterogeneity and suppose \( w_k^M = w_k = 0 \). The aggregate substitutability between foreign and domestic varieties is given by a weighted average of each industry’s corresponding preference parameter. This captures the direct effect of a shock to \( E \) on \( p_{ki}^* \), and the resulting immediate change in \( c_{ki} \) visible from equation (1). In macroeconomic data, traded quantities are summed up to the country level, and the econometrician asks how these aggregates respond to changes in international prices. In other words, she assumes \( \sigma_k = \sigma \) for all \( k \). Under this constraint, \( \eta = 1 - \sigma \) as expected, where \( \sigma \) is the (partial) elasticity of substitution
between domestic and foreign varieties in all sectors. In practice however, $\sigma_k \neq \sigma_s$, $k \neq s$, and the aggregate price elasticity of imports is a biased measure of the aggregate (partial) elasticity of substitution. The magnitude of the bias depends directly on the cross-sector correlation between import shares $n_k$ and their substitutability $\sigma_k$.

In addition, both sectoral and aggregate price indices respond to macroeconomic shocks. Consider a positive shock to the nominal exchange rate, driving up the domestic price of imports, $p_{ki}^*$. The second term in equation (2) captures the response of $P_k$, which by definition increases. This has two effects. First, it mitigates how expensive the domestic variety of good $k$ is relative to its substitutes, since good $k$ as a whole has become dearer. This lowers (in absolute value) the measured (total) elasticity, since it acts to dilute the initial change in $\frac{p_{ki}^*}{P_k}$. Second, it induces substitution away from good $k$, suddenly relatively more pricey, and increases (in absolute value) the measured (total) elasticity since it reinforces the initial shift away from the domestic variety of good $k$. Which effect dominates depends on the sign of $\sigma_k - \gamma$; with higher substitutability between varieties than between goods, the term is positive. In any event, both effects increase with the importance of imports in sector $k$, and the importance of sector $k$ in overall imports.

The third integral in equation (2) reflects the increase in the overall price index $P$, which limits substitutability away from good $k$. This acts to mitigate the response of the quantities demanded of the domestic variety of good $k$, and thus lowers measured total elasticity (in absolute value).

Estimates of the price elasticity of imports obtained from aggregate data provide biased values of the true preference parameter that reflects the aggregate substitutability between domestic and foreign varieties. The bias arises because differences in good-specific substitutability are assumed away. This is true of partial and total elasticity. In what follows we report results for both concepts, correcting for heterogeneity or not. We do so because, in the words of Orcutt (1950), partial elasticities have “generally been estimated for particular commodities, while total have been estimated for all, or large groups, of imports lumped together” (page 118).


3 Identification

We review how the methodology in Feenstra (1994a) is adapted to our purposes. We first discuss the econometrics involved in estimating $\sigma_k$ for all sectors $k$ in the US economy. We emphasize how we accommodate common effects across all sectors and measurement error. We then turn to the estimation of $\sigma$, a measure of elasticity constrained to be identical across sectors. We close with a description of our data.

3.1 Microeconomic Estimates

We identify the substitutability between domestic and foreign varieties using the observed cross-section of traded quantities and prices across exporters to one destination. This is afforded by the crucial assumption of an Armington aggregator between varieties of each good, irrespective of their origin. The assumption is what makes it possible to use Feenstra’s (1994a) methodology in the present context, even though we do not observe prices or quantities of domestically produced varieties. We now describe our implementation of his methodology, but keep the development concise and focused on the modifications we introduce.

Contrary to a large literature in Industrial Organization, the identification of a structural demand parameter is achieved in the absence of any supply-shifting instruments. We now explain succinctly how this obtains. Demand is given in equation (1), which after rearranging writes:

$$c_{kit} = \left( \frac{p^*_{kit}}{P_{kt}} \right)^{1-\sigma_k} \left( \frac{\beta_k P_{kt} - 1}{p^*_{kit}} \right)$$

with $\sigma_k > 0$ the elasticity of substitution between varieties of good $k$. Unlike Feenstra (1994a) or Broda and Weinstein (2006), we are not seeking to compute ideal price indices and so can allow for complementarities, and values of $\sigma_k$ strictly below unity.

Feenstra (1994a) imposes a simple supply structure, and assumes

$$p^*_{kit} = \tau_{kit} E_{it} \exp(u_{kit}) c_{kit}^{\omega_k}$$

where $u_{kit}$ denotes a technological shock that can take different values across sectors and exporters, and $\omega_k$ is the inverse of the price elasticity of supply in sector $k$. This assumes

\[\text{We follow Feenstra (1994a) and assume all exporters have the same supply elasticity.}\]
production decisions are taken on the basis of the price net of transport costs, and labeled in domestic currency.

In practice, the approach uses expenditure shares, not least to alleviate measurement error in unit values following Kemp (1962). We define 

\[ s_{kit} = \frac{p_{kit}^c\mu_{kit}}{P_{kt}\tau_{kit}} \]

and rewrite demand as

\[ s_{kit} = \left( \frac{p_{kit}^c}{P_{kt}} \right)^{1-\sigma_k} \beta_k^{\sigma_k-1} \]

Since the data focus on traded goods only, we do not observe domestically produced consumption. In addition, prices are measured Free on Board but in the importer’s currency. We introduce tilded variables to denote the observed counterparts to theory-implied prices and quantities. We observe \( \tilde{p}_{kit}^* \equiv p_{kit}^* / \tau_{kit} \). The empirical market shares are therefore given by

\[ \tilde{s}_{kit} = \frac{\tilde{p}_{kit}^c\mu_{kit}}{\sum_i \tilde{p}_{kit}^c\mu_{kit}} = s_{kit} \left( 1 + \frac{\tilde{p}_{kit}^c\mu_{kit}}{\sum_i \tilde{p}_{kit}^c\mu_{kit}} \right) = s_{kit} \frac{\mu_{kit}}{\tau_{kit}} \]

Taking logarithms, it is straightforward to rewrite demand as

\[ \Delta \ln \tilde{s}_{kit} = (1 - \sigma_k) \Delta \ln \tilde{p}_{kit}^* + \Phi_{kt} + \varepsilon_{kit} \] (3)

with \( \Phi_{kt} \equiv (\sigma_k - 1) \Delta \ln P_{kt} + \Delta \ln \mu_{kt} \), a time-varying intercept common across all varieties, and \( \varepsilon_{kit} \equiv (\sigma_k - 1) \Delta \ln \beta_{kit} - \sigma_k \Delta \ln \tau_{kit} \) an error term that captures random trade cost and taste shocks via changes in \( \tau_{kit} \) and \( \beta_{kit} \). Feenstra (1994a) shows this implies the demand system is robust to quality changes in variety \( i \) of good \( k \) - or indeed to time-varying number of firms producing good \( k \) in country \( i \). The estimation is robust to the presence of an extensive margin within exporting economies.

After rearranging, substituting in log-linearized supply yields

\[ \Delta \ln \tilde{p}_{kit}^* = \Psi_{kt} + \frac{\omega_k}{1 + \omega_k \sigma_k} \varepsilon_{kit} + \delta_{kit} \] (4)

with \( \Psi_{kt} \equiv \frac{\omega_k}{1 + \omega_k \sigma_k} (\Phi_{kt} + \Delta \ln \sum_i (\tilde{p}_{kit}^c\mu_{kit})) \) a time-varying factor common across varieties, which subsumes sector specific prices and quantities. \( \delta_{kit} \equiv \frac{1}{1 + \omega_k \sigma_k} [\Delta \ln E_{it} + \Delta v_{kit}] \) is an error term encapsulating movements in the exchange rate and technological developments in country \( i \) and sector \( k \). The coefficient estimates are left unaffected by either trade costs or exchange
rates movements, even though they can differ across source countries and sectors. Empirically, prices can indifferently be measured Free on Board or Cost, Insurance, Freight, and in either importer’s or exporter’s currency.

Under standard assumptions on taste shocks $\beta_{kit}$ and technology shocks $\nu_{kit}$, it is possible to identify the system formed by equations (3) and (4). Identification rests on the cross-section of exporters $i$ to the domestic economy, and is achieved in relative terms with respect to a reference country $r$:

$$\begin{align*}
\Delta \ln \tilde{s}_{kit} - \Delta \ln \tilde{s}_{krt} &= (1 - \sigma_k)(\Delta \ln \tilde{p}^*_{kit} - \Delta \ln \tilde{p}^*_{krt}) + \tilde{\varepsilon}_{kit} \\
\Delta \ln \tilde{p}^*_{kit} - \Delta \ln \tilde{p}^*_{krt} &= \frac{\omega_k}{1 + \omega_k \sigma_k} \tilde{\varepsilon}_{kit} + \tilde{\delta}_{kit}
\end{align*}$$

where $\tilde{\varepsilon}_{kit} \equiv \varepsilon_{kit} - \varepsilon_{krt}$ and $\tilde{\delta}_{kit} = \delta_{kit} - \delta_{krt}$. 9 We follow Feenstra (1994a) and summarize the information contained in the system with the following estimable regression

$$Y_{kit} = \theta_{1k} X_{1kit} + \theta_{2k} X_{2kit} + u_{kit} \quad (5)$$

where $Y_{kit} = (\Delta \ln \tilde{p}^*_{kit} - \Delta \ln \tilde{p}^*_{krt})^2$, $X_{1kit} = (\Delta \ln \tilde{s}_{kit} - \Delta \ln \tilde{s}_{krt})^2$, $X_{2kit} = (\Delta \ln \tilde{s}_{kit} - \Delta \ln \tilde{s}_{krt})(\Delta \ln \tilde{p}^*_{kit} - \Delta \ln \tilde{p}^*_{krt})$ and $u_{kit} = \tilde{\varepsilon}_{kit} \tilde{\delta}_{kit} \frac{(\sigma_k - 1)(1 + \omega_k)}{1 + \omega_k \sigma_k}$. Estimates of equation (5) map directly with the parameters of interest, since

$$\begin{align*}
\theta_{1k} &= \frac{\omega_k}{(\sigma_k - 1)(1 + \omega_k)} \\
\theta_{2k} &= \frac{\omega_k \sigma_k - 2\omega_k - 1}{(\sigma_k - 1)(1 + \omega_k)}
\end{align*}$$

Equation (5) still suffers from an endogeneity issue. We follow Feenstra (1994a) and instrument the regressors with country-specific fixed effects, and correct the estimation for heteroskedasticity across exporters $i$. We also include an intercept to account for the measurement error arising from using unit values to approximate prices. Given the origin of potential measurement error, we let it prevail at the most granular level afforded by our data, and allow for intercepts specific to each HS6 sector.10

9 In the empirics, we choose a reference country that is present in the US market during the whole observed period.

10 For the instrumentation to be consistent, there must be some cross-country differences in the relative variance of the demand and supply curves. For an intercept to capture measurement error, its variance must be equal across exporting countries. Of course, it may still be specific to each sector. See Feenstra (1994a, 1994b).
The system summarized by equation (5) can accommodate developments that are specific to each sector $k$. But in macroeconomic applications where the universe of economic activities that form Gross Domestic Product is considered, it is important to allow for more general, aggregate influences. Aggregate technology shocks for instance, presumably affect prices and quantities jointly in all sectors. If it were a shock in the exporting economy, that would correspond to a common component of $u_{kit}$ across all $k$. We allow for such correlated effects in as general and parsimonious a manner as possible. We implement a correction suggested by Pesaran (2006) to purge all “Common Correlated Effects” (CCE) from sector level data, and estimate

$$Y_{kit} = \theta_0 + \theta_{1k} \hat{X}_{1ki} + \theta_{2k} \hat{X}_{2ki} + \theta_{3k} X_{1it} + \theta_{4k} X_{2it} + u_{kit} \tag{6}$$

where the intercept allows for HS6-specific measurement error, hatted variables are the instrumented versions of $X_{1kit}$ and $X_{2kit}$, and $X_{1it}$ and $X_{2it}$ control for the time-varying component of $Y_{kit}$ that is common across all sectors. In particular, following Pesaran (2006), $X_{1it}$ and $X_{2it}$ are the cross-sector arithmetic averages of $X_{1kit}$ and $X_{2kit}$. Equation (6) is corrected for heteroskedasticity across exporters $i$ using the procedure described in Feenstra (1994a).

Armed with consistent (and sector-specific) estimates of $\theta_{1k}$ and $\theta_{2k}$, it is straightforward to infer elasticities. In particular, the model implies

$$\hat{\sigma}_k = 1 + \frac{\hat{\theta}_{2k} + \Delta_k}{2 \hat{\theta}_{1k}} \text{ if } \hat{\theta}_{1k} > 0 \text{ and } \hat{\theta}_{1k} + \hat{\theta}_{2k} < 1$$

$$\hat{\sigma}_k = 1 + \frac{\hat{\theta}_{2k} - \Delta_k}{2 \hat{\theta}_{1k}} \text{ if } \hat{\theta}_{1k} < 0 \text{ and } \hat{\theta}_{1k} + \hat{\theta}_{2k} > 1$$

with $\Delta_k = \sqrt{\hat{\theta}_{2k}^2 + 4 \hat{\theta}_{1k}}$. Appendix A.1 details how these are also used to infer standard deviations around these point estimates.

As is apparent, there are combinations of estimates in equation (6) that do not correspond to any theoretically consistent estimates of $\hat{\sigma}_k$. Even though our universe of acceptable values is slightly broader than Broda and Weinstein (2006) - whose purpose was different - this is a problem we encounter in our data, as they did. We follow their approach, and use a search algorithm that minimizes the sum of squared residuals in equation (6) over the intervals of admissible values of the supply and demand elasticities. We use this approach whenever direct estimates of $\theta_{1k}$ and $\theta_{2k}$ cannot be used to infer $\hat{\sigma}_k$. Whenever CCE are included, we hold
constant the estimates of $\theta_{0k}, \theta_{3k}$ and $\theta_{4k}$ obtained from the direct instrumental variable regression, and search the combination of values for $\sigma_k$ and $\omega_k$ that minimizes the sum of squared residuals in equation (6). The corresponding standard errors are obtained via bootstrapping of the procedure using 1,000 repetitions.

3.2 Macroeconomic Estimates

We invoke equation (2) to aggregate adequately the estimates of $\hat{\sigma}_k$ just obtained. Appendix A.1 details how the corresponding standard deviation is derived, on the basis of a first-order approximation around the true value of the coefficient. This provides the value and confidence intervals for the aggregate parameter of interest, allowing for heterogeneity at the sectoral level. We argue this is a true estimate of the representative agent’s elasticity of substitution between domestic and foreign aggregates, allowing for heterogeneity in substitutability across sectors.

Our purpose is to compare these results to what is obtained when sectoral elasticities are constrained to be homogeneous, as they would in conventional regression analysis based on macroeconomic data. To do so, we impose $\sigma_k = \sigma$ and modify equation (6) into

$$Y_{kit} = \theta_0 + \theta_1 \bar{X}_{1i} + \theta_2 \bar{X}_{2i} + \theta_3 X_{1it} + \theta_4 X_{2it} + u_{kit} \quad (7)$$

We maintain the assumption of a HS6-specific intercept, to continue to accommodate the possibility that measurement error varies at the disaggregated level. Not doing so will conflate two potential sources of bias, and the one we are pursuing relates only to the estimates of $\theta_1$ and $\theta_2$. These are now constrained to equality across all sectors $k$. Identification continues to rest on the cross-section of exporters $i$, but equation (7) is now estimated on the pooled dataset formed by observations on all sectors. It is noteworthy that identification in equations (6) and (7) rests in practice on the same dimension of the same dataset. It is therefore difficult to ascribe the discrepancy we find to a difference in the extent of an extensive margin. Whether our data (and procedure) capture or not firm dynamics, they do so equally in both estimations.

We continue to allow for the possibility that aggregate shocks in any country $i$ should affect all sectors simultaneously, and include adequately modified CCE terms. The instrumentation and correction for heteroskedasticity are also modified accordingly. In particular, country-specific effects are used as instruments across the whole panel of sectors. Equation (7) is
effectively estimated on a panel of HS6 sectors. Our partition of HS6 sectors into 56 ISIC categories is entirely innocuous here. Homogeneity is imposed across all sectors, and two HS6 goods have to be equally substitutable whether or not they belong to the same ISIC sector. This is not true of the heterogeneous estimates, where HS6 goods were constrained to be equally substitutable within an ISIC category - but not between.\footnote{This is the reason why we opt for a panel approach across HS6 sectors, rather than performing a constrained Seemingly Unrelated Regression estimator. The latter is simply impossible across more than 3,000 varieties.}

Armed with estimates of $\theta_1$ and $\theta_2$, it is easy to obtain a value for the constrained elasticity of substitution $\hat{\sigma}$. Our model then implies conventional macroeconomic estimates of the elasticity are given by

$$\hat{\eta}_c = 1 - \hat{\sigma} + (\hat{\sigma} - \gamma) \sum_{k \in K} n_k w_k^M + \gamma \sum_{k \in K} n_k w_k^M w_k$$

with a standard error given by a first-order approximation, as detailed in the Appendix. Following the definitions in Orcutt (1950), $\hat{\eta}_c$ is the constrained total elasticity, and $1 - \hat{\sigma}$ is a constrained partial elasticity.\footnote{$\gamma$ is not estimated in our procedure. Instead, it is arbitrarily fixed to one, the special case of a Cobb-Douglas aggregator at the sectoral level. Calibrating this parameter in a more rigorous way cannot much affect our results, as the term involving $\gamma$ is an order of magnitude smaller than the one involving $\sigma$. In addition, the impact of $\gamma$ is the same in the constrained and unconstrained cases, and therefore cannot matter for the discrepancy between the two.}

### 3.3 Data

The BACI database, issued by CEPII, describes bilateral trade at the sectoral level. The data build on the United Nations ComTrade database, with some added effort put in the harmonization of trade flows on the basis of both import and export declarations. The improvement acts to limit measurement error. The data are defined at the 6-digit level of the harmonized system (HS6) and cover around 5,000 products over the 1996-2004 period. The products are regrouped within industries as defined by the 3-digit level of the ISIC (Revision 3) nomenclature.

The partition into ISIC sectors is only relevant for the unconstrained estimation, and is performed for lack of detailed information on $n_k, w_k^M$ or $w_k$ at such high level of granularity. It corresponds to the assumption that all HS6 goods are equally substitutable within an ISIC
category, but not between. This does presumably assume some heterogeneity away, and possibly creates a bias as a result. We conjecture that heterogeneity between ISIC industries is more sizable, and thus creates more of a bias. We do however perform some robustness in section 4.4, using all HS6 goods in the unconstrained estimation. But to do so we have to maintain some rather stringent assumptions on the values for $n_k$, $w^M_k$ and $w_k$. We note once again that the partition is innocuous for the constrained estimation imposing homogeneity.

The approach adapted from Feenstra (1994a) requires relatively little information on traded flows. To estimate equation (6) we only need measures of $\tilde{p}_{kit}$ and the expenditure shares $\tilde{s}_{kit}$. As is conventional, we use unit values to approximate bilateral prices, and divide values of bilateral trade flows by their volume. In BACI, values are denominated in USD and are Free On Board. Quantities are in tons. The empirical model described in section 3.1 is not sensitive to the currency denomination of trade data, nor to the treatment of trade costs, as both are passed into the residuals. Expenditure shares are measured as $\tilde{s}_{kit} = \frac{\tilde{y}_{kit}^{\text{c}}}{\sum_{i \in I} \tilde{y}_{kit}^{\text{c}}}$.

We subject our data to sampling with a view to limiting the role of extreme outliers. These are notoriously frequent in approaches making use of unit values to approximate prices. For instance, tonnage is not always appropriate to capture the traded volumes of all HS6 goods, which can instill artificial (massive) volatility in the resulting time series on prices. In each sector, we exclude annual variations in prices and market shares that exceed five times the median value. In addition, we impose a minimum of 20 exporters for each HS6 good over the whole observed time period. The cross-section of exporters is what ultimately achieves identification. Measurement error may prevail in estimates of $\hat{\sigma}_k$ that are based on too few exporters, which would translate into biased values for (unconstrained) aggregate elasticities. We require that at least 20 exporters be present to alleviate this concern. Our data ultimately represent 73 percent of the total value of US imports, across 56 ISIC sectors.

The three weights used in aggregation come from a variety of sources. In the main body of the text, we privilege results that correspond to values of $n_k$, $w^M_k$ and $w_k$ taken from as few data sources as possible. In particular, $n_k$ and $w^M_k$ are first computed from the US input/output (IO) tables, available in the ISIC (Revision 3) nomenclature. $n_k$ is defined as the 1997 ratio

\[ \text{In general, trade data are collected by national customs offices in the currency of the declaring country. These data are then converted in US dollars by the United Nations, using the current nominal exchange rate.} \]
of sectoral over total import values, and $w_k^M$ is the 1997 ratio of imports over domestic gross output. In the main body of the text, $w_k$ is taken from the STAN database, as the 1997 ratio of sectoral relative to total interior demand (given by gross output and imports net of exports).

In section 4.4, we verify our results do not depend on this specific choice of data sources. We discuss four alternatives. In all cases, $w_k$ is left unchanged and computed using STAN 1997 data. We focus the sensitivity analysis on alternative measures of $n_k$ and $w_k^M$, because equation (2) suggests $w_k$ has second order impact on estimates of $\eta$. In a first alternative, the import data necessary to compute $n_k$ and $w_k^M$ are taken from BACI (in 1997), while the output data continue to come from the IO tables. An advantage is that the same data are then (partly) used to compute the relevant weights and to estimate the elasticities. In a second alternative, we still use imports from BACI but the output data are taken from STAN, and the ratios use 1999 data.

In these two cases, the values of $w_k^M$ are computed on the basis of two different datasets. In a third alternative, we compute $n_k$ and $w_k^M$ using a single source, $n_k$ from BACI, $w_k^M$ from the I/O tables, all measured in 1997. Finally, up to now, $w_k^M$ is measured using output data, while the theoretical parameter refers to total sectoral consumption. Our fourth and final alternative defines $w_k^M$ as the 1999 ratio of sectoral imports over interior demand, as implied by gross output and imports net of exports, obtained from STAN.

4 Results and Relevance

We first review the microeconomic estimates, obtained across 56 ISIC sectors, and relate them with existing evidence. We then aggregate them, first preserving heterogeneity. We compare the results with our estimation imposing homogeneity - using the same data sources to perform the aggregation. We discuss the discrepancy, and in particular whether its magnitude is significant economically. We argue the corrected estimates we uncover change dramatically the quantitative and qualitative predictions of a vast range of international macroeconomic models. We close with some robustness.
4.1 Microeconomic Results

Figure 1 reports sectoral estimates of \( \hat{\sigma}_k \) for 56 ISIC sectors. On average, \( \hat{\sigma}_k \) is equal to 6.7, with values ranging from 28 to 3.1 and a standard deviation of 4.9. The median value is only 5.1, reflecting a skewed distribution of elasticities: only 5 out of 56 estimates are above 10. How do our results compare with existing studies of the substitutability between foreign and domestic varieties, at similar aggregation levels? If anything, a median value of 5.1 lies at the low end of the range of estimates obtained in the empirical trade literature. Romalis (2007) finds elasticities of substitution between 6.2 and 10.9 at the HS6 level. Head and Ries (2001) find values between 7.9 and 11.4 at the 3-digit SIC level. Hanson (2005) finds estimates between 4.9 and 7.6 using data at the US county level. A common denominator across these studies is their focus on disaggregated, microeconomic information on traded quantities and/or tariffs. In particular, they report cross-sector averages of microeconomic estimates, rather than estimates based on macroeconomic data.

How does the magnitude of our individual sectoral estimates compare with the literature? In theory, the parameters we estimate are comparable with the values obtained in the conventional approach regressing imported quantities on relative prices. The relative price of imports is typically measured with respect to domestically produced varieties, as in Houthakker and
Magee (1969) or Kreinin (1967). The price elasticity of imports is then given by $1 - \hat{\sigma}_k$. These elasticities were the object of a vast literature spread over the 1960s and 1970s, with two differences. First, in most instances, the results put forward focused on short-run elasticity estimates, typically for reasons of data availability especially at a disaggregated level. In contrast, given our assumptions, our estimates correspond to long run elasticities. Second, the data used were more coarse, focusing on just a few sectors. Still, in what follows we strive to ensure our disaggregated estimates are consistent with the existing estimates of import elasticities.

Houthakker and Magee (1969) report in their Table 6 a long run price elasticity in manufactures estimated at -4.05. This is virtually identical to the median value we obtain across our 56 manufacturing sectors, equal to -4.1. Kreinin (1967) documents similar estimates, with an elasticity for manufactures equal to -4.71. It is remarkable that such different data sources, coverages and methodologies should yield strikingly similar median estimates. Our assumptions find comfort in this convergence of results.

The data in Houthakker and Magee (1969) and Kreinin (1967) are much coarser than ours, but they also discuss the relative magnitudes of elasticity estimates across the categories they observe. Manufactures have the higher estimates, followed by semi-manufactures and crude foods and materials. Similar relative rankings come out of the survey in Goldstein and Khan (1985). They summarize their Table 4.4 commenting that “the price elasticity of demand for manufactures is significantly larger than that for non-manufactures. Within non-manufactures, price elasticities for raw materials appear to be larger than those for food and beverages” (pages 1084-1085). A precise mapping is difficult given the differences in granularity, but the ranking is roughly prevalent in our results as well. There are exceptions, but our highest estimates concern finished manufactures, such as aircrafts, TVs, telephones, photo instruments, footwear, motor vehicles or office machinery. At the other end of the spectrum, we find relatively low elasticities for dairy, wood, food, beverages and semi-manufactures like wires or metal products.

Mapping our most disaggregated, sector-specific estimates with the literature becomes quickly difficult, once again because of data availability as of 20 or 30 years ago. In fact, not many papers have attempted to estimate sector-specific price elasticities of imports, say at the two or three digit level of aggregation. We were able to identify three exceptions. Stone
(1979) presents US estimates at the two digit level. On the whole, his estimates are lower than ours, but that can simply reflect his focus on short run elasticities. A few examples may nevertheless help illustrate the relative similarities in our results. For “Inorganic Chemicals”, Stone estimates an import price elasticity of -3.40, as against -3.60 for “Other Chemicals” in Figure 1. He finds -2.22, -2.32, and -3.71 in “Rubber Products”, “Plastic Materials and Articles” and “Dyeing, Tanning and Coloring Agents”, as against -4.1, -3.1 and -5.9 in “Rubber Products”, “Manufactures of Plastic Products” and “Tanning and Dressing of Leather” in Figure 1. Keeping in mind ours are estimates of long-run elasticities, these values lie in similar ballparks.

Shiells (1991) estimates long run elasticities at the three digit SITC level, but only for 12 US sectors. Once again, an accurate mapping is impossible in most cases. Interestingly however, his estimate in “Newsprint” is -3.597, indistinguishable from our value of -3.4 for “Publishing”. He also finds -3.489 in “Steel Plate and Sheet”, not that dissimilar from the estimate of -5.2 we find for “Manufacturing of Basic Iron and Steel”, even though sector definitions are different. The discrepancies become even less substantial when taking into account Shiells’ relatively large standard errors.

Finally, Khan (1975) introduces data from Venezuela, disaggregated at a non-homogeneous, non conventional level, to estimate price elasticities of imports in nine sectors. These include “Food”, with an elasticity of -2.04, “Paper and Cardboard” with value -0.99, or “Textiles” with value -1.67. Keeping in mind these are once again short run elasticities, they are comparable to the estimates we obtain in similar activities. For instance, we get -2.1 in “Manufactures of Dairy Products”, -3.3 for “Manufactures of Beverages”, -3.3 in “Manufactures of Paper Products”, and -3.0 in “Manufactures of Other Textiles”.

This comparison exercise is not meant to suggest we reproduce exactly sector-specific results that were obtained several decades ago in totally different data, using drastically different methodologies, and, in the case of Venezuela, in a different country. Rather, we seek to ascertain the identification strategy we follow is not fundamentally falsified. In particular, the Armington assumption is what exonerates us from having to observe any characteristics of domestically produced goods. With the Armington aggregator, the observed prices and quantities of imports
originating from a cross-section of countries are sufficient to identify the elasticity of interest, between domestic and foreign varieties.

From this point of view, it is reassuring that our mean and median estimates should be strikingly close to seminal, fundamental contributions to the literature on imports price elasticities. Virtually all the papers there do make use of domestic prices in their estimations: import prices are evaluated relative to their domestically produced counterpart. That we should find similar results without any information on domestic prices vindicates the Armington assumption. The few punctual comparisons we report at the level of individual sectors do, as well.

There is of course an obvious comparison that has been absent from our analysis so far. We have implemented a variant of the methodology introduced by Feenstra (1994a), just as Broda and Weinstein (2006) have. Our objectives are fundamentally different, as are some of our identifying restrictions and some of the corrections we introduced. Still, Broda and Weinstein estimate the universe of substitution elasticities in disaggregated US data, just as we do. Given the similarities in methodologies, it is not surprising that our estimates should be similar, and they are. Their mean estimate at the three digit level is 4.0, with a standard deviation of 7.9. “Petroleum Oils and Oils from Bituminous Minerals, Crude”, “Aircraft and Associated Equipment” or “Fuel Wood” are sectors with relatively high elasticities, whereas “Lighting Fixtures”, “Radio-Broadcast Receivers” or “Motorcycles and Cycles” all rank towards the bottom of their list.

But the comparison is not especially informative in terms of validating our assumptions. What is key here is the Armington assumption that the substitutability between two foreign varieties should equal that between domestic and foreign varieties. If this is true in the data, we can infer directly the price elasticity of imports. Since the latter has been the object of a vast empirical literature, it is with it that we have striven to compare our results.

4.2 Macroeconomic Results

We now turn to macroeconomic estimates of the elasticity of substitution implied by the values in Figure 1. To do so, we apply the aggregation procedure spelled out in equation (2). Our main point concerns the difference in estimates of $\eta$ where $\sigma_k = \sigma$ as against those where $\sigma_k$ is
left unconstrained. As is patent, the bias thus induced will increase in the correlations between \( \sigma_k \) and \( n_k \), and between \( \sigma_k \) and \( n_k \cdot w_k^M \). The third term in equation (2) has second order effects only.

Table 1 reports both correlations across the five variants we use to compute \( n_k \) and \( w_k^M \). In all cases, the correlations are mildly positive. This suggests constraining all elasticities to be the same will act to lower estimates of \( \eta \). In fact, we have selected the benchmark weights that tend to minimize the extent of a putative aggregation bias, with lowest values for both correlations. Since the extent of the bias increases in both correlations, Table 1 suggests it will be larger in Variants 2 or 3. The third term in equation (2) can of course slightly affect that ranking, for large enough changes in the correlations between \( n_k, w_k^M, \) and \( w_k \) across variants.

Table 2 reports estimates of the aggregate price elasticity of imports. We first report estimates of \( \eta \), the total elasticity implied by equation (2), when we impose that \( \sigma_k \) be equal across all sectors. Our point estimate suggests a value for the parameter of -1.9, with narrow standard error bands. A confidence interval at standard significance levels implies values ranging roughly from -1.7 to -2.1. This is at the high end of the range of values obtained in conventional estimates of the elasticity based on macroeconomic data. For instance, Goldstein and Kahn (1985) claim that “Harberger’s (1957) judgment of 25 years ago that the price elasticity of import
Table 2: Estimation with common correlated effects

<table>
<thead>
<tr>
<th></th>
<th>Estimate</th>
</tr>
</thead>
<tbody>
<tr>
<td>Constrained total elasticity</td>
<td>-1.916*</td>
</tr>
<tr>
<td></td>
<td>(.119)</td>
</tr>
<tr>
<td>Constrained partial elasticity</td>
<td>-2.738*</td>
</tr>
<tr>
<td></td>
<td>(.048)</td>
</tr>
<tr>
<td>Unconstrained total elasticity</td>
<td>-4.210*</td>
</tr>
<tr>
<td></td>
<td>(.316)</td>
</tr>
<tr>
<td>Unconstrained partial elasticity</td>
<td>-6.006*</td>
</tr>
<tr>
<td></td>
<td>(.481)</td>
</tr>
<tr>
<td>Number of sectors</td>
<td>56</td>
</tr>
<tr>
<td>Number of grid searches</td>
<td>11</td>
</tr>
</tbody>
</table>

Note: Standard errors in parentheses (obtained by bootstrapping for grid searched sectors), * denotes significance at the 1% level.

demand for a typical country lies in or above the range of -0.5 to -1.0 still seems on the mark”. In their Table 4.1, they report estimates for the US between -1.03 and -1.76, an interval that is not significantly distinct from the one we estimate.

Once again, this vindicates the assumption of an Armington aggregator. We obtain an aggregate estimate with nothing but import prices, that is virtually identical to one obtained on the basis of relative prices computed with domestic price indices. More importantly, -1.9 is consistent with the choices made in the vast majority of calibration exercises in international macroeconomics. The implied elasticity of substitution is between two and three, which includes the ranges of values used in, say, Obstfeld and Rogoff (2005) or Backus, Kehoe and Kydland (1994).

This stands in contrast with the value of $\eta$ obtained when $\sigma_k$ is left unconstrained across sectors. As shown in the Table, the parameter jumps to -4.2, with standard errors that guarantee a significant difference at conventional confidence levels. This more than doubles the corresponding value for the estimated elasticity of substitution, at 5.2. We argue this is the value that should enter the utility of a representative agent with heterogeneous preferences.
across sectors. Given the overwhelming evidence that substitutability is heterogeneous across goods or sectors, we contend a value around 5 is preferable from a calibration standpoint. Such a high value characterizes adequately the average substitutability of a representative agent who has heterogeneous preferences across goods.

Why the discrepancy? Does this not mean that aggregate quantities should be measurably more responsive to aggregate relative prices? Orcutt (1950) already speculated an explanation, reported by Goldstein and Khan (1985). On page 1070, they explain that, “in aggregate trade equations, goods with relatively low price elasticities can display the largest variation in prices and therefore exert a dominant effect on the estimated aggregate price elasticity, thereby biasing the estimate downwards.” We view the results in Table 2 as a confirmation of this decade-old conjecture.

Orcutt (1950) introduced the notion of a “partial” elasticity, $\eta_p$, which assumes away aggregate price responses in equation (2). From Section 2, this is simply defined as a weighted average of $\sigma_k$, with weights given by $n_k$. Strictly speaking, this weighted average is the direct equivalent of the aggregate elasticity of substitution, although disaggregated data are necessary to calculate it. We have referred to $1 - \eta$ as the natural proxy for aggregate substitutability, because macroeconomic calibrations would typically infer the elasticity of substitution from the price elasticity of imports, estimated on the basis of macroeconomic data. It is therefore the right comparison with the literature.

It is nevertheless useful to check the aggregation bias we document continues to prevail in estimates of “partial” elasticities $\eta_p$. We report constrained and unconstrained values for $\eta_p$ in Table 2. As is patent, a bias continues to prevail, with a constrained estimate at -2.7, jumping to -6.0 when sector specific elasticities are permitted. The adjustment of price indices in equation (2) does not explain our results away. But as expected, it tends to dilute measured elasticity, as partial elasticities are systematically higher. The Table also says that a “true” estimate of the aggregate elasticity of substitution, purely on the basis of sector specific estimates, and controlling for all adjustment in (sectoral and aggregate) price indices is effectively equal to $1 - (-6.0) = 7.0$. 

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4.3 Relevance

Is the correction we document relevant in economic terms? We now discuss the quantitative and qualitative consequences of our corrected estimates. The most straightforward implication concerns models directly dealing with the responsiveness of traded quantities to relative prices, and in particular the resolution of “global imbalances”. Most prominently, Obstfeld and Rogoff (2005) use a calibrated model to argue a reversal of the US current account is compatible with a 30% depreciation of the real exchange rate. The calibration sets substitutability at 2. In a slightly simplified two-country version, we obtained depreciation rates of 22 and 21% for values of the parameter of 5 and 7, respectively, down from 31% with an elasticity of 2.14 The parameter is quantitatively important, and shaves off two-thirds of the “required” depreciation, almost all the way to the 19.3% that obtains for an elasticity of 100. This is true even though Obstfeld and Rogoff’s calibration gives prominence to another parameter, the elasticity of substitution between traded and non-traded goods, important in this instance because the US is a largely closed economy. Still, the effects are sizeable and probably important in terms of welfare as well.

Cole and Obstfeld (1991) showed the endogenous response of the terms of trade can deliver perfect insurance against country-specific shocks when the elasticity of substitution between domestic and foreign goods is exactly unitary. The result is meant as an illustrative special case of a powerful mechanism. Still, models of international portfolio holdings will have drastically different qualitative predictions depending on which side of one the parameter lies. For instance, Heathcote and Perri (2008) show that, with complete markets, imperfect substitutability between domestic and foreign goods can generate a home equity bias. A positive domestic productivity shock will increase the relative return on domestic stocks as long as the terms of trade do not respond too strongly: there will be portfolio home bias for values of the elasticity of substitution below one.

A contrario, in Coeurdacier (2005), domestic consumers choose to hold foreign assets to insure against shocks to domestic consumption, provided international relative prices (the terms of trade) respond strongly enough in response to shocks in the relative quantities produced.

14 We are grateful to Cedric Tille for graciously giving us the simulation code.
This happens for values of the elasticity above one. Both papers then move on to introduce labor income risk and endogenous production, or incomplete markets. Our purpose here is not to settle the question of the origins of an equity home bias. Rather, we stress the fact our estimated aggregate elasticity is substantially above one has far-reaching implications in models of international portfolio choice.

Empirically, trade partners display highly correlated business cycles. Burstein, Kurz and Tesar (forthcoming) show this is quantitatively compatible with a model with production sharing. A positive domestic technology shock diffuses internationally because it increases the demand for foreign inputs, so that both trade and cycle correlations are high simultaneously. A key ingredient is the strong complementarity between the domestic and foreign goods involved in domestic production. Vertical linkages in production are absent from our analysis. Still, our universally high elasticity estimates suggest such strong complementarities are absent in sectoral data (though they may well prevail in highly specific within-firm trade flows). This is part of what gives rise to a co-movement puzzle in Kose and Yi (2006).

Atkeson and Burstein (forthcoming) propose to explain observed deviations from PPP in a model with trade costs, imperfect competition and variable markups. Variables profit margins require that the elasticity of substitution between the (foreign and domestic) varieties that form a sector be larger than the substitutability across sectors. In their model, markups then depend on market shares, and can therefore fluctuate over time. In their calibration, the elasticity equals 10; deviations from PPP virtually disappear for an alternative value set at 3. Mirdigan (forthcoming) uses disaggregated data from Eastern European countries to conclude menu costs are insufficient to explain the pattern of international relative prices. High trade costs or low elasticities of substitution are also needed. On average at least, our results suggest trade costs are of the essence.

It is perhaps not surprising that models of the real exchange rate should have predictions that depend on the substitutability between domestic and foreign goods. A direct corollary is that the policy consequences of international price differences will also depend crucially on the parameter. Presumably, the relevance of the exchange rate in the monetary policy rule developed in Galí and Monacelli (2005) depends on the substitutability between foreign and
domestic varieties. Galí and Monacelli focus on unitary elasticity, so the result is not directly apparent there. But Benigno and De Paoli (2006) introduce a generalization of their model, and their conclusions point to that direction. As in Cole and Obstfeld (1991) with unitary elasticity, a marginal reduction in the utility value of output is accompanied by an exactly corresponding reduction in the utility value of consumption. This insulates the economy from terms-of-trade movements. With non unitary elasticity however, policy shocks that affect the terms-of-trade also affect welfare, in a way that crucially depends on whether the calibrated parameter is above or below one. The fact that we find an aggregate estimate substantially above one must therefore have important policy implications.

4.4 Stability

This section verifies the robustness of our results in three dimensions. First, we ascertain our results do not depend on a particular choice of data source in computing the crucial weights defined in equation (2). Second, we investigate the importance of “Common Correlated Effects” in obtaining estimates of $\sigma_k$, and ultimately of $\eta$. Third, we relax our assumption that elasticities of substitution be identical across the HS6 categories regrouped in each ISIC sector. Instead, like Broda and Weinstein (2006) we estimate a value of $\sigma_k$ for each HS6 category. We discuss the necessary shortcuts this requires in terms of aggregation.

Table 3 compares the constrained and unconstrained total import elasticities $\eta$ using different weighting vectors. The data sources and computations behind the four alternative variants we present in the Table are discussed in Section 3.3. The first line repeats the results implied by the benchmark weights we have used so far. As expected given the empirical values for the correlations between $\sigma_k$ and $n_k$, and between $\sigma_k$ and $n_k \cdot w^M_k$ reported in Table 1, the benchmark weights give rise to the smallest heterogeneity bias. Across the four variants, constrained estimates of $\eta$ are as low as -1.8, and unconstrained estimates almost reach 5. Variant 3 gives rise to the largest discrepancy, a factor of 2.5. The bias continues to be quantitatively important across these four alternatives.

The inclusion of Common Correlated Effects in the estimation of $\sigma_k$ is justified by our interest in the macroeconomic implications of the microeconomic values we obtain. After all, the
Table 3: Variants on the weights

<table>
<thead>
<tr>
<th></th>
<th>Total elasticity</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Unconstrained</td>
<td>Constrained</td>
<td></td>
</tr>
<tr>
<td>Benchmark</td>
<td>-4.210</td>
<td>-1.916</td>
<td></td>
</tr>
<tr>
<td>Variant 1</td>
<td>-4.872</td>
<td>-2.083</td>
<td></td>
</tr>
<tr>
<td>Variant 2</td>
<td>-4.422</td>
<td>-1.981</td>
<td></td>
</tr>
<tr>
<td>Variant 3</td>
<td>-4.987</td>
<td>-1.969</td>
<td></td>
</tr>
<tr>
<td>Variant 4</td>
<td>-4.157</td>
<td>-1.814</td>
<td></td>
</tr>
</tbody>
</table>

Note: Benchmark: $n_k$ and $w_k^M$ using imports and output from IO tables, $w_k$ using STAN sectoral interior demand. Variant 1: $n_k$ and $w_k^M$ using imports from BACI and output from STAN, $w_k$ using STAN sectoral interior demand. Variant 2: $n_k$ and $w_k^M$ using imports from BACI and output from STAN, $w_k$ using STAN sectoral interior demand. Variant 3: $n_k$ using imports from BACI, $w_k^M$ using imports and output from IO tables, $w_k$ using STAN sectoral interior demand. Variant 4: $n_k$ using imports from BACI, $w_k^M$ using imports and interior demand from STAN, $w_k$ using STAN sectoral interior demand.

quantities traded at sector level, and their prices, do presumably respond to common, aggregate, macroeconomic influences. When estimating the “true”, sector-specific substitutability between domestic and foreign varieties, one want to ascertain one is not capturing aggregate dynamics. This would amount to double-counting at the time of aggregation. Does this matter in our estimations? Table 4 provides a mixed answer, using again our benchmark weights. Without CCE, constrained estimates of $\eta$ increase slightly, to -2.11, whereas unconstrained estimates decrease slightly, to -3.83. These changes are not strongly significant relative to our benchmark results, and they do not alter the conclusion of a significant heterogeneity bias. But they nevertheless suggest the introduction of a CCE term in equation (6) is not innocuous.

Finally, we relax our assumption that the substitutability between two HS6 categories be identical within each ISIC sector. In other words, we allow for heterogeneity in $\sigma_k$ even within each ISIC sector. Like Broda and Weinstein (2006), we estimate an elasticity of substitution for each HS6 sector, and then use equation (2) to aggregate them at the macroeconomic level. This raises the question of what values for $n_k$, $w_k^M$ and $w_k$ to use: we do not observe any direct information on any of these weights at such a refined aggregation level. We choose to impose similar values of the weights for all HS6 categories that belong to one ISIC sector. Clearly, this
Table 4: Estimation without common correlated effects

<table>
<thead>
<tr>
<th></th>
<th>Estimate</th>
</tr>
</thead>
<tbody>
<tr>
<td>Unconstrained total elasticity</td>
<td>-3.831&lt;sup&gt;a&lt;/sup&gt;</td>
</tr>
<tr>
<td></td>
<td>(.122)</td>
</tr>
<tr>
<td>Unconstrained partial elasticity</td>
<td>-5.524&lt;sup&gt;a&lt;/sup&gt;</td>
</tr>
<tr>
<td></td>
<td>(.206)</td>
</tr>
<tr>
<td>Constrained total elasticity</td>
<td>-2.111&lt;sup&gt;a&lt;/sup&gt;</td>
</tr>
<tr>
<td></td>
<td>(.027)</td>
</tr>
<tr>
<td>Constrained partial elasticity</td>
<td>-3.016&lt;sup&gt;a&lt;/sup&gt;</td>
</tr>
<tr>
<td></td>
<td>(.041)</td>
</tr>
<tr>
<td>Number of sectors</td>
<td>56</td>
</tr>
<tr>
<td>Number of grid searches</td>
<td>12</td>
</tr>
</tbody>
</table>

Note: Standard errors in parentheses (obtained by bootstrapping for grid searched sectors), *a* denoting significance at the 1% level.

assumes away some possible source of a heterogeneity bias, but there is simply no alternative. But we know choosing other values for \( n_k \) and \( w_k^M \) do not matter at the ISIC level. At least, this tests whether the heterogeneity in estimates of \( \sigma_k \) within each ISIC category can be such that our conclusions are altered.

Naturally, the constrained estimates of \( \eta \) continue to be identical, for instance at -2.11 without a CCE corrective term. After all, this is an estimation that constrains all coefficients to be identical, within and between HS6 categories. The difference arises for unconstrained estimates. We estimated values for \( \sigma_k \) in 4,021 HS6 categories, and aggregated them using our benchmark (ISIC) weights. We obtained a value of -4.84 for the unconstrained elasticity. Computing standard error bands across this point estimate is not tractable, but the value is virtually identical to Variants 1 and 3 in Table 3, and less than 0.6 away from the others. We conclude ignoring heterogeneity within ISIC sectors is not important to our conclusions.
5 Conclusion

The elasticity of substitution between domestic and foreign varieties is central in international economics. But no clear consensus has emerged from a vast empirical literature seeking to pin down the parameter, except for one essential disagreement. On average, microeconomic data tend to imply substantially higher values than macroeconomic aggregates. The point estimates are also much more heterogeneous. We propose that this heterogeneity is the reason why aggregate results are close to zero. We compute structural estimates of aggregate substitutability allowing or not for heterogeneity at the sectoral level. We find that imposing homogeneity is enough to obtain aggregate estimates in line with the macroeconomic evidence, even using a disaggregated dataset. Allowing for heterogeneity results in an aggregate parameter value of up to 5. This discrepancy validates the conjecture of an aggregation bias in elasticity estimates that goes back at least to Orcutt (1950). Such high parameter values change dramatically the conclusions of calibrated models in areas of international economics as varied as the international transmission of shocks, global imbalances, international risk sharing, portfolio choice and optimal monetary policy.

References


Coeurdacier, N., 2005, “Do trade costs in goods market lead to home bias in equities?”, mimeo.


A.1 Estimated Variances

The variance of \( \hat{\sigma}_k \) is computed using the second-order moments of \( \hat{\theta}_{1k} \) and \( \hat{\theta}_{2k} \) and a first-order approximation of \( \sigma_k \) around its true value:

\[
\sigma_k = \hat{\sigma}_k + \left. \frac{\partial \sigma_k}{\partial \theta_{1k}} \right|_{\theta_{1k} = \hat{\theta}_{1k}} (\hat{\theta}_{1k} - \hat{\theta}_{1k}) + \left. \frac{\partial \sigma_k}{\partial \theta_{2k}} \right|_{\theta_{2k} = \hat{\theta}_{2k}} (\hat{\theta}_{2k} - \hat{\theta}_{2k})
\]

\[
\Rightarrow Var(\hat{\sigma}_k) = \left( \left. \frac{\partial \sigma_k}{\partial \theta_{1k}} \right|_{\theta_{1k} = \hat{\theta}_{1k}} \right)^2 Var(\hat{\theta}_{1k}) + 2 \left( \left. \frac{\partial \sigma_k}{\partial \theta_{1k}} \right|_{\theta_{1k} = \hat{\theta}_{1k}} \right) \left. \frac{\partial \sigma_k}{\partial \theta_{2k}} \right|_{\theta_{2k} = \hat{\theta}_{2k}} Cov(\hat{\theta}_{1k}, \hat{\theta}_{2k}) + \left( \left. \frac{\partial \sigma_k}{\partial \theta_{2k}} \right|_{\theta_{2k} = \hat{\theta}_{2k}} \right)^2 Var(\hat{\theta}_{2k})
\]

where:

\[
\begin{align*}
\left. \frac{\partial \sigma_k}{\partial \theta_{1k}} \right|_{\theta_{1k} = \hat{\theta}_{1k}} &= \frac{1}{\hat{\theta}_{1k}} \left[ 1 - \sigma + \frac{1}{\sqrt{\hat{\theta}_{2k}^2 + 4\hat{\theta}_{1k}}} \right] \\
\left. \frac{\partial \sigma_k}{\partial \theta_{2k}} \right|_{\theta_{2k} = \hat{\theta}_{2k}} &= \frac{1}{2\hat{\theta}_{1k}} \left[ 1 + \frac{\theta_{2k}}{\sqrt{\hat{\theta}_{2k}^2 + 4\hat{\theta}_{1k}}} \right]
\end{align*}
\]

Using the same reasoning, the first-order approximation of the aggregate elasticity around its estimated value gives:

\[
\eta = \hat{\eta} - \sum_{k \in K} n_k (1 - w_k^M)(\sigma_k - \hat{\sigma}_k)
\]

The variance of \( \hat{\eta} \) is then defined as:

\[
\Rightarrow Var(\hat{\eta}) = \sum_{k \in K} n_k^2 (1 - w_k^M)^2 Var(\hat{\sigma}_k) + \sum_{k \in K} \sum_{k' \neq k} n_k n_{k'} (1 - w_k^M)(1 - w_{k'}^M) Cov(\hat{\sigma}_k, \hat{\sigma}_{k'})
\]

In the constrained case, the same reasoning gives

\[
Var(\hat{\eta}^c) = \left[ \sum_{k \in K} n_k (1 - w_k^M) \right]^2 Var(\hat{\sigma})
\]