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TRADE, INSECURITY, AND HOME BIAS: AN EMPIRICAL INVESTIGATION

James E. Anderson Douglas Marcouiller

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ABSTRACT

Corruption and imperfect contract enforcement dramatically reduce trade. This paper estimates the reduction, using a structural model of import demand in which transactions costs impose a price markup on traded goods. We find that inadequate institutions constrain trade far more than tariffs do. We also find that omitting indexes of institutional quality from the model leads to an underestimate of home bias. Using a broad sample of countries, we find that the traded goods expenditure share declines significantly as income per capita rises, other things equal. Cross-country variation in the effectiveness of institutions offers a simple explanation of the observed global pattern of trade, in which high-income, capital-abundant countries trade disproportionately with one another.

James E. Anderson Department of Economics Boston College Chestnut Hill, MA 02167 and NBER james.anderson@bc.edu

Douglas Marcouiller Department of Economics Boston College Chestnut Hill, MA 02167 douglas.marcouiller@bc.edu Recent findings have re-emphasized an old puzzle: far too much of the variation in trade volume across country pairs is accounted for by distance and border effects (Frankel, Stein and Wei 1998; Helliwell 1998; McCallum 1995). Distance is more important than can plausibly be explained by transportation costs (Grossman 1998; but also see Hummels 1998). Borders matter far more than can be explained by tariffs, quotas, and formal impediments to trade. Informal impediments and hidden transactions costs may explain these results. Understanding such hidden costs would improve policy design and stimulate progress in theoretical models of trade and development. This paper focuses on insecurity as one source of hidden costs.

Anecdotal evidence of the importance of insecurity abounds. A survey undertaken by the World Bank between August 1996 and January 1997 summarizes such stories well. Table 1 shows the ranking in order of importance of "the obstacles for doing business," based on responses by 3685 firms located in 69 countries. It is not surprising that firms should complain about taxes; it is remarkable, however, that corruption should rank as the second most import obstacle to business worldwide, with crime and theft not far behind. Complaints about trade regulations, currency and price controls, and labor and environmental regulations appear relatively insignificant.

This paper sets out a structural model of insecure trade and fits it to 1996 trade flows. Our results show not only *that* insecurity matters, but how much it matters and to whom it matters. Our structural modeling approach offers some insight into *why* insecurity matters.

We model two types of insecurity, one arising from predation (theft, corruption, extortion) and the other arising from imperfect contract enforcement. Each is shown to imply a price markup analogous to a hidden tax on trade. When predation takes the form of theft, the markup is determined by the probability that a particular shipment will be hijacked. When predation takes the form of bribes, the markup is equal to the proportion of the value of each shipment which shippers expect to lose. These are equivalent when risks can be diversified through insurance or by making a large number of small shipments subject to independent risks. Imperfect contract enforcement leads to a slightly different markup. When entry into the international market involves sunk costs, imperfect contract enforcement exposes shippers to the holdup problem even when the investment is not partner-specific. In this case, the exogenous probability of enforcement determines the size of the price markup.

	Worldwide Sample
Tax Regulations or High Taxes	1
Corruption	2
Financing	3
Inadequate Infrastructure	4
Crime and Theft	5
Inflation	6
Uncertainty of Cost of Regulations	7
Policy Instability	8
Labor Regulations	9
Regulations on Foreign Trade	10
Safety or Environmental Regulations	11
Start-up Regulations	12
Foreign Currency Regulations	13
Price Controls	14
Terrorism	15

Table 1. Rankings of "Obstacles for Doing Business"

Source: Brunetti, Kisunko, and Weder, 1997, p. 70.

Using data on institutional quality compiled by the World Economic Forum, we show that trade expands dramatically when it is supported by strong institutions – specifically, by a legal system capable of enforcing commercial contracts and by transparent and impartial formulation and implementation of government economic policy. We estimate, for example, that if the indexes of institutional quality associated with the Latin American countries in our sample were to improve to the levels associated with the European Union, Latin American trade would expand by 34%, other things equal, outweighing the projected impact of a move to free trade. Attention to the costs of insecurity may help in solving Trefler's (1995) "mystery of the missing trade" in embodied factor services.

We show that empirical models which ignore the security of exchange suffer from an important omitted variables bias. We find very significant "home bias," an income elasticity of demand much less than one, when institutional indicators enter the model; one calculation shows that the share of total expenditure devoted to traded goods falls 3.5% when per capita income rises 10%. In contrast, when security is ignored, no home bias is found; the positive impact of strong institutions is misattributed to high income per capita, the included variable with which institutional quality is correlated. The home bias effect reported here stands in contrast to recent empirical work which has failed to reject homothetic preferences (Davis and Weinstein 1998; Davis, Weinstein, Bradford and Shimpo 1997). Our work leads us to echo Trefler (1995, p.1043), "the bias is important and must be confronted theoretically and empirically."

The stylized fact that high-income capital-abundant countries trade disproportionately with each other rather than with low-income laborabundant countries has been used to motivate models based on product differentiation rather than factor endowments, but insecurity provides an alternative explanation. The price effect of good institutional support for trade among high-income countries leads them to trade disproportionately with one another. This argument does not imply, counterfactually, that lowincome countries should also trade disproportionately with one another.

The paper begins by showing how insecurity translates into a price markup. The second section ties the price markup into import demand. The third section describes the data which are used to estimate the model in the fourth section. The fifth sections reports several checks on the robustness of our results.

1. Modeling the Security of Trade

Two types of insecurity can generate price markups equivalent to a hidden tax on international trade. A model of "predation" views shipments as subject to attack by hijackers or corrupt officials. A model of "contractual insecurity" captures the impact of the holdup problem on shippers when fixed costs are associated with entry into the international market and contract enforcement is random. These are complementary rather than competing models. Each leads to a simple price markup which is a reduced form function of exogenous variables. Either model can motivate the demand system estimated later in this paper.

Predation

Anderson and Marcouiller (1998) present a complete general equilibrium model of predation, in which utility-maximizing agents rationally allocate their labor across productive and predatory activities, endogenously determining the probability of successful shipment. Here we present a slightly simplified version of the model.

Thieves – or corrupt officials – attack shipments. Any shipment which is defended by less than the customary measures is identifiable as easy prey, attacked with certainty, and completely lost. Under these circumstances, all shippers will take the normal defensive measures and thieves will attack randomly. The probability that a normally defended shipment from country *i* will be lost is given by the asymmetric contest success function:

(1.1)
$$_{i} = \frac{1}{1 + \frac{L_{i}^{B}}{L_{i}^{D}}},$$

a function of total labor devoted to banditry L_i^B , total labor devoted to defense L_i^D , and an exogenous technological parameter $.^1$ The ability to diversify risk makes equivalent, from the shippers' point of view, to a proportional

insecurity tax of on the value of every shipment. This tax is bounded on the unit interval, increasing in bandit labor and decreasing in defensive labor.

In this paper we treat defensive arrangements L_i^D as given.² We also assume the world's total supply of bandit labor to be exogenously set: $L^B = \int_i L_i^B$. The endogenous allocation of bandits across countries then determines \int_i . Bandits freely allocate themselves across countries in a competitive equilibrium so as to maximize expected loot $\int_i \left(1 - \int_i \left(L_i^B, \overline{L}_i^D, \cdot\right)\right) v_i$, where v_i is the volume of trade flowing through the border of country *i*. The reasonable assumption that uncoordinated bandits take trade volumes as given greatly simplifies this problem. Solving the first order conditions gives the allocation of bandit labor to each country:

(1.2)
$$L_i^B = \frac{i(1-i)v_i}{i(1-i)v_i}L^B.$$

A bit of algebra produces the reduced form solution for *i*:

(1.3)
$$_{i} = \frac{L_{i}^{D}}{V_{i}} \frac{^{1/2}}{\frac{L^{B}}{V_{i}} + w_{i}(L_{i}^{D} / v_{i})^{1/2}}}{\frac{L^{B}}{V_{i}} + w_{i}(L_{i}^{D} / v_{i})}$$

where w_i is country i's share of total world trade.

Let $\bar{S}_i = (L_i^D / v_i)^{1/2}$ denote the strength of a country's institutions for the defense of trade. Then:

(1.4)
$$_{i} = \overline{S}_{i} \qquad w_{i}\overline{S}_{i} \qquad / \frac{L^{B}}{V_{i}} + w_{i}\overline{S}_{i}^{2}$$
.

If the probability of successfully crossing into country j is independent of the probability of successfully leaving i, the proportion of all shipments from

¹ The same function has been used in the context of non-anonymous predation by Grossman and Kim (1995).

² This is, of course, a major simplification. See Anderson and Marcouiller (1998).

producers in *j* which successfully reach their consumers in *i* is given by:

(1.5)
$$_{ij} = _{i j} = \overline{S}_i \overline{S}_j \qquad W_i \overline{S}_i^2 / \frac{L^B}{V_i} + _{i} W_i \overline{S}_i^2$$
.

The probability of loss on this trade route, $(1 - i_j)$, determines the transactions cost and the corresponding price markup associated with insecurity.

Equation 1.5 can be extended to include other influences on $_{ij}$. When the two countries share a common border (represented by a dummy b_{ij}) or a common language (dummy l_{ij}), $_i$ and $_j$ may not be independent. The risk of theft might rise as the distance traveled rises, perhaps due to loss of information about ways to avoid hazards.³ Adding these variables and changing to the considerably simpler relative security form $_{ij} \swarrow_{kj}$ produces the equation:

(1.6)
$$\frac{j}{k_j} = \frac{\overline{S}_i}{\overline{S}_j} \frac{1+b_{ij}}{1+b_{kj}} \int_{1}^{1} \frac{1+l_{ij}}{1+l_{kj}} d_{ij}^{3}$$

The price markup on imports by country *i* from country *j* relative to the markup on imports by *k* from *j* will reflect the relative probability of successful shipment, as described in Section 2 below.

Contractual Insecurity

Insecurity in the form of imperfect contract enforcement generates a price markup when fixed costs are associated with entry into the international market. Following Anderson and Young (1999), we model a market in which for institutional reasons there is some exogenous probability (1 -) that a given contract may fail to be enforced. When contracts are not enforced, the contracting parties engage in *ex post* bargaining, in which the sunk costs of trade (all handling charges up to the point of sale) are ignored. Foreseeing this

possibility, high cost traders are discouraged from entering the market. The effect on trade can be modeled as a price markup equivalent to a tariff. The sketch of the model we present here is necessarily cursory, serving only to give the elements which yield a plausible reduced form which we take to our empirical work. See Anderson and Young (1999) for details.

Sunk costs are associated with entry into international trade.⁴ International exchange occurs either according to the terms of a contract negotiated prior to incurring the sunk costs or in a non-contracted market into which those whose contracts are not enforced necessarily fall. We allow traders without enforced contracts to match only once per trading period.⁵ In the non-contracted market, exchange occurs at the bargained price

$$p^* = \underset{p}{\operatorname{argmax}}(p-c) (b-p)^{1-} = b + (1 -)c$$

where *b* and *c* are the exogenously determined outside options (home prices) for the buyer and seller and (0,1) is the bargaining strength of the seller.

In these circumstances, it is only by accident that the numbers of buyers and sellers would be equal. Any unmatched trader will return home to exchange at his outside option price. We focus in this development on the excess demand case, in which some potential importers are unable to find exporters to deal with.

The actual volume exchanged is that on the short side of the market, read off the supply curve, $s[p^s(p^*, ,b)]$, where p^s is the equilibrium value of the certainty equivalent price to suppliers, which can be shown to be a reduced form function of the bargained price, the probability of enforcement

³ This is the only point at which we mention information costs, but we do not wish to deny their importance. For a provocative model of information costs and trade, see Casella and Rauch (1998).

⁴ In the usual holdup model, these costs are relationship-specific: the exporter designs a product for a particular importer. The outside option of the exporter is whatever resale value this design has for others. Similarly the outside option of the importer is whatever price must be paid for an equivalent design elsewhere. Here, we need not assume that the sunk costs are relationship specific because we assume that search is so expensive that traders match only once.

and the outside option of the buyers. To obtain the "tariff equivalent" of the imperfect enforcement we first define the hypothetical buyers' price which would clear the market at the actual trade volume:

 $p^{t}(p^{*}, , b) = \{p \mid d[p] = s[p^{s}(p^{*}, , b)].$

Then the ad valorem tariff equivalent is

(1.7)
$$T(p^*, , b) = \frac{p^t(p^*, , b)}{p^s(p^*, , b)} - 1$$

The ad valorem tariff equivalent is decreasing in , (see Anderson and Young, 1999) hence better enforcement increases trade.

In our application, the assumed exogenous varies across countries so that country *j*'s exports face different markups in each country *i*. The p* and b arguments of T() are handled as follows. The bargained price p* is a weighted average of the sellers' and buyers' reservation prices. The seller's reservation price is set at unity by convention and is invariant across buyers. The buyers' reservation price *b* is modeled as a reduced form function of exogenous endowment variables. Finally, the weights in the bargained price are assumed to be equal for all country pairs, because in the absence of a bargaining theory which can discriminate among countries, it seems best to assume that 1- is the same across buyers. Under these assumptions, the security questionnaire data we use as proxies for accurately pick up the effect of differing security arrangements on price markups.

2. Import Demand in an Insecure World

The strength of a nation's institutions affects the prices it must pay for traded goods, as shown in the previous section. Import demand depends in

⁵ If rematching were possible, the trader who is faced with returning home to his outside option could offer a better deal than the bargained price to someone about to accept the bargain. That is, the outside option would be endogenous.

turn on these prices and on the division of expenditure between traded and non-traded goods.

Our model of import demand assumes two-stage budgeting. Agents first determine the proportion of total expenditure to allocate to traded goods. In a second stage they allocate traded goods expenditure across individual imports, which are differentiated by place of origin.⁶ The first-stage preferences are not restricted. Preferences across traded goods are CES and identical across countries.

Under these assumptions, the impact of prices on demand in country *i* for imports from country *j* is given by:

(2.1) $m_{ij} = {}_{j}p_{ij}^{-}P_{i}^{-1}x_{i}$ where x_{i} is country i's total expenditure on traded goods, p_{ij} is the price of j's good in *i* with producer prices p_{jj} normalized to one, $P_{i} = {}_{j} {}_{j}p_{ij}^{1-}$ is the CES price index for traded goods in *i*, is the elasticity of substitution among traded goods, and ${}_{j}$ is that parametric expenditure share on j's product

which is common to all importers.

The country's total expenditure on traded goods, x_i , is some fraction of the country's total income. The traded goods expenditure share is modeled as a reduced form function of the country's income, population and traded goods price index. A variety of static structural models yield such a function.⁷ Anderson (1979) rationalized this reduced form with a model of perfect competition and constant returns to scale. Bergstrand (1985, 1989) developed the reduced form from a model with monopolistic competition and economies of scale. The equilibrium price of the nontraded good is a reduced

⁶ Helliwell 1998, p. 10. notes other papers using this Armington assumption.

⁷ Our empirical work explains trade in a single year, so static models are appropriate. In reality, balanced trade is rare and the traded goods expenditure share reflects an intertemporal margin of decision-making. We ignore this margin because it is remote from the concerns of our model and seems unlikely to add to its explanatory power. Temporary trade control measures taken for balance of payments reasons will show up in the traded goods price index.

form function in the same variables and is subsumed in the traded goods expenditure share function. Income and population pick up the effect of factor endowments, possible nonhomothetic preferences and possible scale economies, while the traded goods price index picks up substitution between traded and nontraded goods. Substituting into 2.1:

(2.2)
$$m_{ij} = {}_{j}p_{ij}^{-}P_{i}^{-1}(y_{i}, n_{i}, P_{i})y_{i}$$

where n_i is population and y_i is national income.

Insecurity enters the model through its effect on prices. The price of *j* 's product in *i* will exceed the producer's price for three reasons: a tariff if applicable, a transport cost dependent on distance, and an "insurance" markup which captures either the proportion of shipments lost to predators $(1 - _{ij})$ or the tariff-equivalent markup attributable to insecure enforcement of contracts $\frac{p^t(\ ,b,p^*)}{p^s(\ ,b,p^*)}$ –1 . In both models of international insecurity, p_{ij} decreases and m_{ij} increases as the effectiveness of institutions for the defense of exchange improves.

Three additional simplifications have proven enormously helpful in moving toward an estimable model. First, we use loglinear approximations of the basic functions. We approximate the price markup as a log-linear function of distance, security, and the tariff factor, if applicable. If instead transportation and insurance markups are modeled additively, the model becomes deeply nonlinear. We also model the reduced form function as loglinear.

Second, we focus on m_{ij} / m_{kj} , country *i*'s imports from country *j* relative to country *k*'s imports from country *j*, instead of looking at m_{ij} directly. This makes the model invariant to multiplicative rescaling of the WEF data, and it allows us to cancel some of the nonlinear terms of the $_{ij}$ function.

More importantly, casting the model in terms of relative imports by two different countries from a single exporter eliminates the need to estimate the $_j$ parameter. Empirical models following Anderson's (1979) rationale for the gravity equation are usually misspecified. The gravity model is derived from the import demand system by imposing the adding up constraint that shipments to the entire world be equal to income, solving that constraint for the expenditure share for each exporter and finally substituting the exporterspecific expenditure share into the import demand equation. Anderson shows that the correct specification of the gravity equation includes a highly nonlinear exporter-specific price index on the right hand side. Nonlinear structural estimation might be possible, but failing this, an exporter-specific intercept is indicated. Unfortunately, such an intercept cannot be identified since the model also requires other exporter-specific independent variables. Focusing on imports by *i* and *k* from the same exporter *j* eliminates this problem.

Imposing loglinearity on the price markup and using the results of the previous section, Equation 2.2 implies:

(2.3)
$$\ln \frac{m_{ij}}{m_{kj}} = - {}_{1}\ln \frac{d_{ij}}{d_{kj}} - {}_{2}\ln \frac{\overline{S}_{i}}{\overline{S}_{k}} - {}_{3}\ln \frac{1+b_{ij}}{1+b_{kj}} - {}_{4}\ln \frac{1+l_{ij}}{1+l_{kj}} - {}_{4}\ln \frac{1+l_{ij}}{1+l_{kj}} - {}_{4}\ln \frac{1+l_{ij}}{1+l_{kj}} - {}_{4}\ln \frac{1+l_{ij}}{1+l_{kj}} + {}_{6}\ln \frac{1+l_{ij}}{1+l_{kj}} - {}_{4}\ln \frac{1+l_{kj}}{1+l_{kj}} - {}$$

where a_{ij} is a dummy variable which takes the value one if the two countries are associated in a free trade agreement and t_i is the importer's average ad valorem tariff. Through its effect on relative prices, a rise in the contract model's relative probability of enforcement, $i \neq k$, would have an effect similar to that of a rise in the predation model's relative defensive capacity, $\bar{S}_i \neq \bar{S}_k$.

Our third simplifying move is to approximate the relative traded goods price indexes by a version of the Törnqvist index:

(2.4)
$$\ln \frac{P_i}{P_k} = w_j \ln \frac{p_{ij}}{p_{kj}}$$

where w_j is the average across importers of the share of *j*s product in import expenditures. Most previous work with gravity-type models has ignored the price index term, which certainly results in misspecification. Our approximation is an imperfect but sensible and operational measure.

All the major elements of our model are now in place. We have modeled a world in which traded goods are differentiated by place of origin. Differences across importers in demand for a single good have two sources: (a) differences in the price markups associated with insecurity, distance, and tariffs, and (b) differences in the division of expenditure between traded and nontraded goods.

3. Data

The security of transactions depends upon the institutions which structure interaction among private firms and between private firms and the state. We rely on data provided by the World Economic Forum (WEF) to measure the quality of both sets of institutions. The measures are drawn from the WEF 1997 Executive Survey, which was completed by more than 3000 participants distributed across 58 countries (World Economic Forum 1997, p.85). Participants in the WEF survey were asked to assign a score ranging from one (strongly disagree) to seven (strongly agree) to each of the following statements:

- Government economic policies are impartial and transparent (Q 2.07);
- The legal system in your country is effective in enforcing commercial contracts (Q 8.06).

We rescale the mean response for each country to run from zero to one and use the rescaled means as measures of institutional quality, understanding Question 2.07 to gauge primarily the quality of interaction of the private sector with the state and Question 8.06 to gauge institutional support for exchange within the private sector.

These are noisy signals of institutional strength. Expectations differ across countries, so that what counts as "effective" enforcement or "impartial" policy in the Ukraine may differ from what would be similarly classified in Singapore. The respondents to the survey form a selected group – even if they were randomly selected within a country, they would still represent only those who had chosen not to relocate or to shut down. Moreover, the Forum provides only the mean response for each country; we lack information about within-country variation in responses.

Our data on 1996 bilateral import volumes are taken from the IMF's Direction of Trade Statistics.⁸ Data on 1996 population and GDP in current U.S. dollars are taken from the World Bank's World Development Indicators (WDI); since trade flows are measured in current dollars, GDP is measured in the same units. We calculate distance from capital city to capital city on the basis of geographical coordinates listed in Fitzpatrick and Modlin (1986), although the distance from Washington to Ottawa only roughly captures the average distance traversed by shipments from the United States to Canada. David Tarr and Francis Ng of the World Bank graciously provided us with unweighted average external tariff data; these data are far more complete than the data on import duties as a percentage of import expenditures offered by the WDI.⁹ We composed dummy variables to capture sharing a common border, a common language, or common membership in ASEAN, the EU, Mercosur, or NAFTA.

We have complete data on these variables for a total of 2206 import flows distributed across 48 importing countries. In 29 of these 2206 cases, zero

⁸ These statistics are generally reported cif. Eight of the countries in our sample report imports fob. In seven of those cases we have adjusted the flows by the cif/fob ratio of the importer's trade with the world as a whole, as reported in the Direction of Trade Statistics. That ratio was not reported for the Czech Republic, whose fob import figures we adjusted upward by a factor of 1.1.

⁹ Even so, not every country has data available for 1996. We have used 1996 data where available, but in other years have used tariff data from 1997, 1995, or 1994.

imports were reported. Table 2 shows the importing countries in our data set and the number of import flows which we have for each.

IMPORTER	Obs.	IMPORTER	Obs.	IMPORTER	Obs.
Argentina	46	Hungary	47	Russia	47
Australia	47	Iceland	46	Singapore	46
Austria	47	India	47	Slovak Republic	47
Belgium-Luxembourg	47	Indonesia	46	South Africa	47
Brazil	47	Ireland	47	Spain	47
Canada	47	Italy	47	Sweden	47
Chile	41	Japan	47	Switzerland	46
China	47	Jordan	45	Thailand	44
China: Hong Kong	47	Korea	37	Turkey	47
Colombia	46	Malaysia	46	Ukraine	44
Czech Republic	47	Mexico	39	United Kingdom	47
Denmark	47	Netherlands	47	United States	47
Egypt	47	New Zealand	47	Venezuela	45
Finland	47	Norway	46	Zimbabwe	42
France	47	Peru	45		
Germany	47	Poland	47		
Greece	46	Portugal	47	Total	2206

Table 2. Importers in the Data Set

4. Estimation and Results

Our analytical model leads to a simple result – relative import demand is a function of relative income, relative population, relative distance, relative tariffs, and the variables associated with relative security. Estimation of the log-linear model by OLS with robust standard errors supports three contentions:

- By lowering transactions costs, institutional support for secure exchange significantly raises international trade volume;
- Excluding institutional variables biases the estimated coefficient on income upward, understating "home bias" in goods trade;
- The institutional differences which we model can generate "a disproportionately high volume of trade among high-income countries," a pattern "which happens to accord well with trade patterns in the real

world" (Deardorff 1998, p.16).

Equation 2.3 and the Törnqvist index described by Equation 2.4 give us the following model in terms of the underlying parameters:

$$\ln \frac{m_{ij}}{m_{kj}} = (1 + 1) \ln \frac{y_i}{y_k} + 2 \ln \frac{n_i}{n_k} - 1 \ln \frac{d_{ij}}{d_{kj}} + (3 - 1) 2 \ln \frac{s_{1i}}{s_{1k}}$$

$$+ (3 - 1) 2 \ln \frac{s_{2i}}{s_{2k}} - 3 \ln \frac{1 + b_{ij}}{1 + b_{kj}} - 4 \ln \frac{1 + l_{ij}}{1 + l_{kj}}$$

$$+ (3 - 1) 5 \ln \frac{1 + (1 - a_{ij})t_i}{1 + (1 - a_{kj})t_k} + (-1 + 3) 1 \int_{j} w_j \ln \frac{d_{ij}}{d_{kj}}$$

$$+ (-1 + 3) \int_{j} w_j \ln \frac{1 + b_{ij}}{1 + b_{kj}} + (-1 + 3) \int_{j} w_j \ln \frac{1 + l_{ij}}{1 + l_{kj}}$$

Both dimensions of institutional quality are included, assuming that the "defensive capacity" variable of the predation model, \bar{S}_i , involves both private-private and private-public transactions:

$$\frac{\overline{S}_i}{\overline{S}_k} = \frac{S_{1i}}{S_{1k}} \frac{1}{S_{2i}} \frac{S_{2i}}{S_{2k}}$$

The indicators of institutional quality do not vary across exporters for a single importer; the "weighted average" institutional terms collapse into the unweighted terms. Therefore, the coefficient on each institutional index includes its effect on the price, _2 , the direct effect of the price on imports, - , and the indirect effect of the price through the price index, $(-1 + _3)$. The "weighted average" tariff markup is nearly identical to the unweighted tariff markup, since few of the 2130 observations involve free trade, and these two terms have also been combined. This leaves us with:

$$\ln \frac{m_{ij}}{m_{kj}} = {}_{0} + {}_{1} \ln \frac{y_{i}}{y_{k}} + {}_{2} \ln \frac{n_{i}}{n_{k}} + {}_{3} \ln \frac{d_{ij}}{d_{kj}} + {}_{4} \ln \frac{s_{1i}}{s_{1k}} + {}_{5} \ln \frac{s_{2i}}{s_{2k}}$$

$$(4.2) + {}_{6} \ln \frac{1 + b_{ij}}{1 + b_{kj}} + {}_{7} \ln \frac{1 + l_{ij}}{1 + l_{kj}} + {}_{8} \ln \frac{1 + (1 - a_{ij})t_{i}}{1 + (1 - a_{kj})t_{k}} + {}_{9} \int_{j} w_{j} \ln \frac{d_{ij}}{d_{kj}}$$

$$+ {}_{10} \int_{j} w_{j} \ln \frac{1 + b_{ij}}{1 + b_{kj}} + {}_{11} \int_{j} w_{j} \ln \frac{1 + l_{ij}}{1 + l_{kj}} + {}_{ij}$$

as the equation to estimate.

Table 3 reports summary statistics for the import, income, population, distance, transparency, enforceability, adjacency, language, and tariff ratios, as defined above, using the USA as a convenient base country k. Robustness of the results with respect to the choice of the base is explored below.

Ratio: USA as Base	Number Observations	Mean	Standard Deviation
Import Ratio	2130	0.280	0.958
Income Ratio	2130	0.080	0.173
Population Ratio	2130	0.333	0.828
Distance Ratio	2130	1.205	1.850
Transparency Ratio	2130	1.085	0.370
Enforceability Ratio	2130	0.833	0.226
Common Border Ratio	2130	1.026	0.238
Common Language Ratio	2130	0.948	0.263
Tariff Ratio	2130	1.035	0.068

Table 3. Ratios with USA as Base Country

Table 4 reports the results of estimating Equation 4.2 under various restrictions. Results in the first three columns reflect OLS estimation with robust standard errors. Since 29 bilateral trade flows were reported as zero, either because they actually were zero or because they fell below a reporting threshold, we also estimated a tobit version of the model.¹⁰ The fourth column presents the tobit results.¹¹

As shown in the first column, both of the institutional quality

¹⁰ With an elasticity of substitution among traded goods which exceeds one, high transactions costs can eliminate trade in some bilateral pairings.

¹¹ In this case, the value -12.2 was assigned as the log of the import ratio, $\ln(0)$. This value is slightly below the log of the lowest positive import ratio in the data set.

variables have positive and significant coefficients. A few examples shed light on the magnitude of the implied effects. The enforceability of commercial contracts is rated roughly 10% higher in Belgium than in Brazil. Interpreting the estimated coefficient as a reduced form elasticity, this difference in enforceability implies 4% higher imports into Belgium than into Brazil, other things equal.

Variable	OLS 1	OLS 2	OLS 3	Tobit
Log Income Ratio	0.653	0.962	0.881	0.663
	(0.051)	(0.026)	(0.039)	(0.058)
Log Population Ratio	0.197	-0.132	-0.037	0.245
	(0.056)	(0.029)	(0.044)	(0.061)
Log Distance Ratio	-1.106	-1.138	-1.119	-1.135
	(0.040)	(0.038)	(0.039)	(0.043)
Log Transparency Ratio	0.548			0.591
	(0.095)			(0.108)
Log Enforceability Ratio	0.407			0.447
	(0.124)			(0.138)
Log Border Ratio	0.761	0.891	0.803	0.685
	(0.171)	(0.169)	(0.168)	(0.200)
Log Language Ratio	0.324	0.311	0.321	0.326
	(0.090)	(0.091)	(0.091)	(0.115)
Log Tariff Ratio	-4.139		-2.304	-4.488
	(0.879)		(0.837)	(0.958)
Weighted Log Distance Ratio	0.371	0.413	0.415	0.282
	(0.080)	(0.080)	(0.080)	(0.096)
Weighted Log Border Ratio	-0.944	-1.609	-1.499	-0.948
	(0.644)	(0.638)	(0.638)	(0.925)
Weighted Log Language Ratio	-0.263	1.171	1.215	0.379
	(0.616)	(0.593)	(0.585)	(0.825)
Constant	-0.232	0.004	0.020	-0.169
	(0.071)	(0.068)	(0.068)	(0.107)
Number Observations	2130	2130	2130	2159
R-squared	.70	.69	.69	
Log Likelihood				-3933

Table 4. Relative Import Demand, USA as the Base

Robust standard error in parentheses.

The mean enforceability rating among the twelve countries at the low end of the distribution is 0.52 (relative, as always, to the rating of the USA). The mean enforceability rating among the twelve countries in the highest quartile of the distribution is 1.08. A country which saw the measure of the enforceability of its commercial contracts decline from 1.08 to .52 would see its import volume decline 35%, other things equal.¹²

The elasticity of import demand with respect to the transparency and impartiality of economic policy is even higher. Other things equal, imports into France should be on average about 5.5% higher than imports into Argentina simply because the transparency rating is about 10% higher in France than in Argentina.

Taking both institutional indicators into account simultaneously, if the seven Latin American countries in our sample (Argentina, Brazil, Chile, Colombia, Mexico, Peru, and Venezuela) were to enjoy the same transparency and enforceability scores as the mean ratings of the members of the European Union, the predicted ratio of Latin American import volumes to those of the USA would rise from .054 to .072, an increase of 34%.¹³ A 57% increase in average Latin American GDP would be necessary to generate a comparable increase in imports.¹⁴ Moving the seven Latin American countries in the sample to free trade would raise their average imports by only 30%.¹⁵

These calculations take into account both the direct effect of insecurity on the "insurance" markup and the substitution effects associated with the change in price. The price index has been included in a simple and easily operationalized way. The signs of the parameter estimates are plausible (implying, with reference to Equation 4.1, that $_{3} < 1$ and $_{+3} > 1$), and in the case of distance-related price effects remarkably significant. Our model of the impact of the price index on imports implies that $_{3}/_{9}$, $_{6}/_{10}$, and $_{7}/_{11}$ should all be equal. An F-test on the estimated coefficients does not reject that hypothesis.¹⁶

¹⁶ The F-statistic for the joint hypothesis that $\frac{1}{3} / \frac{1}{9} = \frac{1}{6} / \frac{1}{10}$ and $\frac{1}{6} / \frac{1}{10} = \frac{1}{7} / \frac{1}{11}$ is F(2,2118)=2.67.

¹² The projected decline in the log of the import ratio is $0.41^{*}(\ln(1.08)-\ln(0.52))$.

¹³ The projected rise in the log of the import ratio as both institutional ratings rise to EU levels is: $0.41*(\ln(.98)-\ln(.62)) + .55*(\ln(1.19)-\ln(.98))$.

¹⁴ The .30 rise in log relative imports requires an increase of .65*ln(1.57) attributable to relative income.

¹⁵ The projected rise in the log of the import ratio for the tariff decrease is 4.14*(ln(1.065)).

These results signal an enormous impact of institutional quality on trade volume. In fact, in the contemporary world poor institutions appear to constrain trade more dramatically than do tariffs. The estimates justify our first and most important conclusion: by lowering transactions costs, institutional support for secure exchange significantly raises international trade volume.

Our second major finding is that higher income per capita significantly reduces the share of expenditure devoted to traded goods, all else equal. Results in the first column of Table 4 imply that a 10% rise in income per capita will raise imports by only 6.5%; there is quantitatively and statistically significant home goods bias. This result stands in sharp contrast to previous results in the gravity model literature, where traded goods expenditure shares are nearly invariant to income, because the omission by the earlier literature of institutional variables correlated with per capita income biases upward the estimated income effect. Comparison across the columns of Table 4 will reveal the bias clearly.

The share of income spent on traded goods appears in the model as the reduced form function $(y_i, n_i, P_i) = y_i \cdot n_i \cdot P_i \cdot a_i = (y_i / n_i) \cdot n_i \cdot a_i + a_i \cdot P_i \cdot a_i$, with total expenditure on traded goods given by y_i . Therefore, if \hat{f}_i is the coefficient on the log of the income ratio in Table 3, the elasticity of the relative trade expenditure share with respect to changes in per capita income, \hat{f}_i , is given by $\hat{f}_i = 1$. Inappropriately excluding the institutional and tariff variables from the model leads to the results shown in the second column of Table 4, where income per capita has no statistically significant impact on the trade expenditure share ($\hat{f}_i = -1 = -.04$, standard error .03). This is the usual result. When the tariff variable is included but the security variables are not, as in the third column of Table 4, income per capita has a small but significant negative effect on the trade expenditure share ($\hat{f}_i = -1 = -.12$, standard error .04). When the complete model is estimated, as in the first column of results, the elasticity of the trade share with respect to per capita income is

1 - 1 = -.35, with standard error of .05.

The source of the bias is correlation between the omitted variables and income per capita. There is a fairly strong negative correlation between relative income per capita and the tariff ratio (-.6). When the tariff term was excluded, part of its negative effect was incorrectly read as a positive effect of income, so that excluding tariffs from the model led to an underestimate of home bias. The data show a positive correlation between income per capita and the transparency ratio (.5) and between income per capita and the enforceability ratio (.7).¹⁷ Excluding these variables incorrectly attributes to income part of the increase in trade which accompanies the traded goods price reduction associated with good institutions, again underestimating home bias.

Omission of these variables does not significantly bias the estimate of the "size effect," if population is taken as a measure of size. In our model, size affects imports through the traded goods expenditure share:

 $(y_i, n_i, P_i) = (y_i / n_i)^{+} n_i^{-1+} P_i^{-3}$. The elasticity of imports per capita with respect to country size is given by $_1 + _2 -1$. This is estimated using the results of Table 4 as $\hat{}_1 + \hat{}_2 -1$. The model excluding tariffs and institutional quality estimates the elasticity of imports per capita with respect to country size as -17, the model with tariffs negative -16, and the full model -15, all with standard error of .02.

Our home bias result --- other things equal, doubling per capita income reduces the traded goods expenditure share by 35% --- implies a very significant departure from homotheticity. This stands in contrast to the most recent applied trade literature (Davis and Weinstein, 1998; Davis, Weinstein, Bradford and Shimpo, 1997). We coincide with Trefler (1995) in identifying the importance of home bias but diverge from him in tying home bias to

¹⁷ This correlation is given in the data, but it does not imply that income per capita and institutional quality are necessarily linked, nor does it invalidate the "thought experiment" reported above in which institutions were improved without a corresponding increase in income per capita.

income per capita; Trefler uses income per capita as an indicator of factoraugmenting technological differences across countries. Our aggregate results using the reduced form trade expenditure share bear some resemblance to earlier disaggregated work by Hunter and Markusen (1988).

Of course, our model recognizes that the negative effect of income per capita on the trade expenditure share could in practice be offset by a price effect, since the better institutions and lower tariffs of the high-income countries lower the traded goods price index. Combined income and price effects explain why the data show a small positive correlation (.13) between income per capita and total imports divided by GDP.¹⁸

Our final contention is that institutional differences can generate "a disproportionately high volume of trade among high-income countries," a pattern "which happens to accord well with trade patterns in the real world" (Deardorff 1998, p.16). Why should high-income countries skew their trade toward imports from other high-income countries – in spite of the presumed similarity of factor endowment? And what answer to the first question can be consistent with the stylized fact that low-income countries do *not* rely disproportionately on imports from other low-income countries?

Several solutions to the puzzle have been proposed (notably Markusen 1986). We offer an explanation based on the price markup associated with insecure trade. Effective institutions in the importing country lower transactions costs, lower the prices of traded goods, and raise imports, holding constant the characteristics of the exporting country. The predation model argues that the complete price markup also depends on the quality of institutions in the exporting country. Our empirical results confirm that low security in country *i* lowers m_{ij} / m_{kj} ; the predation model also implies that both m_{ij} and m_{kj} are low when the security of country *j* is low. We cannot estimate this second effect, because the impact of the exporter's security and of

¹⁸ The ratio of imports to GDP is not an exact measure of the traded goods expenditure share. It excludes expenditure on the domestically produced tradable good and includes expenditure on goods which are re-exported.

the expenditure share _j are not separately identified. The prediction of the model, however, clearly coincides with the observed pattern of trade. Trade among high-income countries with high-quality institutions ought to be high because the transactions costs associated with insecurity are low; transactions costs impose a double disadvantage on trade among low-income, low-security countries. This solves a problem alluded to in Deardorff's (1998, p.16) informal exposition of an explanation based on identical but non-homothetic preferences. Our story implies disproportionate trade among consumers of the "high-income" good, but it does not imply (counterfactually) disproportionate trade among low-income consumers.

5. Robustness

In this section we briefly examine four questions: How do the estimated parameters differ when the base country is changed? How do they differ when different indexes of institutional strength are used? How do they differ when lagged GDP is used as an instrument for current GDP? Can more general functional forms be estimated?

In theory, there is no reason to suspect that the change of the base country *k* would make any difference to the parameter estimates. In fact, we run into two problems. We have no data on home consumption of the exported good. Therefore, for any base country *k*, we lack a measure of m_{kk} . Since we have no denominator for the relative import measure m_{ik} / m_{kk} , we can never include any country's imports from the base country in the sample used in estimation. Results could be sensitive to the exclusion of differing sets of 47 import observations. A second problem is tied to measurement error. Many of our independent variables take the form $\ln(x_i / x_k)$. The measurement error associated with x_k depends on the choice of *k*, so the parameter estimates may vary with the choice of the base country.¹⁹

¹⁹ This is also a loose justification for allowing an intercept.

Table 5 presents the results of estimating the model with the USA, Brazil, and Japan as alternative base countries. As always, these are OLS estimates of the model with robust standard errors. Given expected difficulties, the results are remarkably consistent across base cases. This is particularly true for the coefficients on the key institutional indexes. For the transparency and enforceability ratios, the coefficients estimated using Brazil and Japan as the bases are well within one standard error of the coefficients estimated using the USA as the base.

Variable	USA Base	Brazil Base	Japan Base
Log Income Ratio	0.653	0.662	0.710
	(0.051)	(0.045)	(0.050)
Log Population Ratio	0.197	0.179	0.119
	(0.056)	(0.048)	(0.055)
Log Distance Ratio	-1.106	-0.978	-1.126
	(0.040)	(0.031)	(0.031)
Log Transparency Ratio	0.548	0.550	0.492
	(0.095)	(0.083)	(0.097)
Log Enforceability Ratio	0.407	0.401	0.343
	(0.124)	(0.110)	(0.130)
Log Border Ratio	0.761	0.939	0.704
	(0.171)	(0.164)	(0.228)
Log Language Ratio	0.324	1.108	0.423
	(0.090)	(0.103)	(0.148)
Log Tariff Ratio	-4.139	-4.163	-3.073
	(0.879)	(0.732)	(0.865)
Weighted Log Distance Ratio	0.371	0.292	0.466
	(0.080)	(0.067)	(0.075)
Weighted Log Border Ratio	-0.944	-0.778	-0.604
	(0.644)	(0.589)	(0.621)
Weighted Log Language Ratio	-0.263	-0.704	-0.204
	(0.616)	(0.345)	(0.395)
Constant	-0.232	0.383	0.266
	(0.071)	(0.051)	(0.075)
Number Observations	2130	2130	2130
R-squared	.70	.73	.68

Table 5. Relative Import Demand, Alternative Base Countries

Robust standard error in parentheses.

Would we find similar results if alternative measures of institutional quality were used? Our variables were chosen to reflect institutions

facilitating exchange within the private sector and between the private sector and the state. The World Economic Forum's survey also asked respondents to score their agreement or disagreement with the statement:

• Irregular additional payments are uncommon in business and official transactions (Q 8.03).

This single statement attempts to capture the both private-private and private-public interactions.

Shang-jin Wei has obtained from the WEF the individual respondents' answers to this question, and he has shared with us the within-country standard deviation of the responses. High within-country variance in individual perceptions of corruption may indicate arbitrariness; if all respondents agree that "irregular payments" are very frequent, then the bribe may be treated as a known cost of doing business. The standard deviation of the response is an indication of the uncertainty in corruption, which may have an independent effect on business activity. Adjusting these statistics to run from 0 to 1, as before, and noting that the *infrequency* of bribes ought to have a positive effect on imports while the *arbitrariness* should have a negative effect, we re-estimated Equation 4.2 using the USA as the base case. Table 6 gives the summary statistics on the new variables:

Ratio: USA as Base	Number Observations	Mean	Standard Deviation
Infrequency of Bribes Ratio	2130	0.732	0.264
Arbitrariness Ratio	2130	1.024	0.360

 Table 6. Additional Ratios with USA as Base Country

Using the logs of these ratios and estimating 4.2 again leads to the results given in Table 7. The middle column reports the result of substituting the bribe-based measures for the transparency and enforceability measures. The coefficient estimate on each has the expected sign, but the coefficient on bribes has a low t-statistic (1.5). This may reflect the relatively low standard deviation of the bribe ratio across countries. It may reflect the high correlation between the bribe ratio and income per capita (.8). It may reflect

measurement error. Imprecision in capturing the security effect may well be responsible for an upward bias on the income coefficient.

Log Income Ratio 0.653 0.786 0.669 Log Population Ratio 0.197 0.046 0.175 (0.056) (0.053) (0.057) Log Distance Ratio -1.106 -1.121 -1.103 (0.040) (0.039) (0.040) Log Transparency Ratio 0.467 . 0.465 (0.095) . (0.121) . 0.405 Log Enforceability Ratio 0.407 . 0.455 (0.124) . (0.132) . Log Infrequency of Bribes Ratio . 0.181 -0.316 Log Border Ratio Log Language Ratio 0.324 0.320 0.329 . . Log Sorder Ratio Log Language Ratio </th <th>Variable</th> <th></th> <th></th> <th></th>	Variable			
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Log Income Ratio	0.653	0.786	0.669
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$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Log Population Ratio	0.197	0.046	0.175
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$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Log Border Ratio	0.761	0.776	0.760
Log Language Ratio 0.324 0.320 0.329 (0.090) (0.091) (0.090) Log Tariff Ratio -4.139 -2.572 -4.415 (0.879) (0.839) (0.890) Weighted Log Distance Ratio 0.371 0.537 0.344 (0.080) (0.085) (0.090) Weighted Log Border Ratio -0.944 -0.615 -0.762 (0.644) (0.686) (0.695) 0.298 Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) 0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) 0.130 Number Observations 2130 2130 2130 R-squared .70 .69 .70		(0.171)	(0.168)	(0.173)
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Log Language Ratio	0.324	0.320	0.329
Log Tariff Ratio -4.139 -2.572 -4.415 (0.879) (0.839) (0.890) Weighted Log Distance Ratio 0.371 0.537 0.344 (0.080) (0.085) (0.090) Weighted Log Border Ratio -0.944 -0.615 -0.762 (0.644) (0.686) (0.695) Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 2130 R-squared .70 .69 .70		(0.090)	(0.091)	(0.090)
$\begin{array}{ccccccc} & (0.879) & (0.839) & (0.890) \\ & & (0.879) & (0.839) & (0.890) \\ & & (0.371 & 0.537 & 0.344 \\ & & (0.080) & (0.085) & (0.090) \\ & & (0.090) \\ & & (0.640) & (0.686) & (0.695) \\ & & (0.644) & (0.686) & (0.695) \\ & & (0.616) & (0.638) & (0.650) \\ & & (0.616) & (0.638) & (0.650) \\ & & (0.071) & (0.068) & (0.074) \\ & & (0.074) \\ & & & Number Observations & 2130 & 2130 & 2130 \\ & & R-squared & .70 & .69 & .70 \\ \end{array}$	Log Tariff Ratio	-4.139	-2.572	-4.415
Weighted Log Distance Ratio 0.371 0.537 0.344 (0.080) (0.085) (0.090) Weighted Log Border Ratio -0.944 -0.615 -0.762 (0.644) (0.686) (0.695) Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 2130 R-squared .70 .69 .70		(0.879)	(0.839)	(0.890)
(0.080) (0.085) (0.090) Weighted Log Border Ratio -0.944 -0.615 -0.762 (0.644) (0.686) (0.695) Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 2130 R-squared .70 .69 .70	Weighted Log Distance Ratio	0.371	0.537	0.344
Weighted Log Border Ratio -0.944 -0.615 -0.762 (0.644) (0.686) (0.695) Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 R-squared .70 .69 .70		(0.080)	(0.085)	(0.090)
(0.644) (0.686) (0.695) Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 R-squared .70 .69 .70	Weighted Log Border Ratio	-0.944	-0.615	-0.762
Weighted Log Language Ratio -0.263 0.121 0.298 (0.616) (0.638) (0.650) Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 R-squared .70 .69 .70		(0.644)	(0.686)	(0.695)
(0.616)(0.638)(0.650)Constant-0.232-0.104-0.306(0.071)(0.068)(0.074)Number Observations213021302130R-squared.70.69.70	Weighted Log Language Ratio	-0.263	0.121	0.298
Constant -0.232 -0.104 -0.306 (0.071) (0.068) (0.074) Number Observations 2130 2130 R-squared .70 .69 .70		(0.616)	(0.638)	(0.650)
(0.071)(0.068)(0.074)Number Observations21302130R-squared.70.69.70	Constant	-0.232	-0.104	-0.306
Number Observations 2130 2130 2130 R-squared .70 .69 .70		(0.071)	(0.068)	(0.074)
R-squared .70 .69 .70	Number Observations	2130	2130	2130
	R-squared	.70	.69	.70

Table 7. Alternative Indexes of Institutional Quality: USA Base

The final column of Table 7 suggests that the bribery question is simply not the best indicator of institutional quality. When all four measures of institutional quality are used together, neither the infrequency nor the arbitrariness of bribes is strongly significant (and the infrequency variable has, if anything, the wrong sign). The other coefficients are much as they were before. This reinforces our confidence that the results presented in the earlier section are reliable.

Variable	Base Results	Instrumented
		Income Ratio
Log Income Ratio	0.655	0.630
	(0.052)	(0.051)
Log Population Ratio	0.193	0.215
	(0.057)	(0.057)
Log Distance Ratio	-1.111	-1.110
	(0.041)	(0.041)
Log Transparency Ratio	0.555	0.594
	(0.097)	(0.097)
Log Enforceability Ratio	0.393	0.433
	(0.127)	(0.126)
Log Border Ratio	0.810	0.801
	(0.182)	(0.182)
Log Language Ratio	0.331	0.332
	(0.092)	(0.093)
Log Tariff Ratio	-4.102	-4.253
	(0.907)	(0.910)
Weighted Log Distance Ratio	0.355	0.360
	(0.081)	(0.081)
Weighted Log Border Ratio	-1.009	-0.926
	(0.668)	(0.670)
Weighted Log Language Ratio	-0.222	-0.350
	(0.636)	(0.637)
Constant	-0.248	-0.274
	(0.074)	(0.073)
Number Observations	2037	2037
R-squared	.68	.68

 Table 8. Relative Import Demand, USA as the Base

Robust standard error in parentheses.

A full general equilibrium model of the economy would treat GDP as endogenous, perhaps with trade encouraging growth. This suggests that correlation between the GDP regressor and the error term of the import regression may have led us to biased parameter estimates. Therefore, we reestimated Equation 4.2 using lagged GDP as an instrument for current GDP.²⁰ The results, which exclude German trade due to a data problem, are presented in Table 8. The first column is our usual specification, the second uses lagged GDP. The new parameter estimates are well within one standard error of the old and strengthen, if anything, the security and home bias effects.

A final point concerning robustness of the results: we experimented with more general functional forms. We tried a translog specification of

defensive capacity instead of using $\frac{\overline{S}_i}{\overline{S}_k} = \frac{s_{1i}}{s_{1k}} \left(\frac{s_{2i}}{s_{2k}} \right)^2$. A Wald test could not reject the hypotheses that the coefficients on all the second order terms were jointly zero, so we returned to the log-linear specification. We also tried to estimate a translog as an approximation to the trade share function

 $\frac{(y_i, n_i, P_i)}{(y_k, n_k, P_k)}$ but found that we could not identify all the necessary parameters

with information on 47 countries.

6. Summary and Conclusion

Abundant anecdotal evidence suggests that transactions costs associated with insecure exchange significantly impede international trade. Predation by thieves or by corrupt officials generates a price markup equivalent to a hidden tax or tariff. Insecure enforcement of contracts can have the same effect. These price markups significantly constrain international trade where legal systems poorly enforce commercial contracts and where economic policy lacks transparency and impartiality.

This paper builds a structural model of import demand in an insecure world and estimates that model using data collected by the World Economic Forum. We find that a 10% rise in a country's index of transparency and impartiality leads to a 5% increase in its import volumes, other things equal.

²⁰ More precisely, using data from World Development Indicators, we multiplied the figure for 1995 GDP in current local currency units by the ratio of the country's 1996 GDP deflator to its 1995 GDP deflator and converted that result to 1996 dollars using the official exchange rate.

A 10% rise in the index of enforceability of commercial contracts leads to a 4% increase in import volume. These estimates are robust with respect to the choice of the base country. Significant costs are associated with institutional weakness. They beg for serious consideration as we try to solve "the mystery of the missing trade" (Trefler, 1995).

We find that the share of total expenditure devoted to traded goods declines significantly as income per capita rises, other things equal. This result stands in sharp contrast to the frequent practice of using homothetic preferences in trade models and to recent findings that homothetic preferences cannot be rejected by statistical tests. The latter finding is replicated here when tariffs and the institutional variables are excluded. Based on this, we claim that omitted variable bias accounts for others' failure to reject homotheticity. The home bias effect of higher income tends to be counterbalanced by a decline in the price index of traded goods as income per capita rises, so that there is in the end a small positive correlation between income per capita and import expenditure as a share of GDP.

Finally, the paper suggests an explanation for the stylized fact that highincome, capital-abundant countries trade disproportionately with each other. These countries are also, in our data, the countries with strong institutions for the defense of exchange. Since the traded goods price markup depends on the degree of insecurity in both the exporting and the importing countries, trade among the rich countries will be relatively unhampered by securityrelated transactions costs, while trade among poor countries will be doubly disadvantaged.

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