

The Globalization Risk Premium[†]

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Abstract

We investigate how the displacement risk associated with import competition is reflected in the cost of capital. We use shipping costs to measure the vulnerability of U.S. industries to import competition. We find that output and employment in high exposure industries is more sensitive to tariff cuts than in low exposure industries, consistent with the idea that they face a higher risk of being displaced by import competition. We then show that high exposure industries have a higher cost of capital. We confirm displacement risk of import competition is priced and covaries with the marginal utility of the representative agent.

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1 Introduction

The dramatic increase in import penetration is among the most important changes that affected the U.S. economy over the past decades. The share of imports in the consumption of manufacturing goods in the U.S. has increased almost fivefold between 1975 and 2005, reaching 25%¹. This fact has attracted a lot of scrutiny but its implications are still debated. Among the benefits of increased import competition are the availability of more product variety at lower prices (Broda and Weinstein, 2006); in addition, recent evidence suggests that domestic firms respond to the threat of import competition by investing in innovation (Bloom et al., 2011; Amiti and Khandelwal, 2013). On the other hand, a stream of research has emphasized the adverse consequences of import competition for U.S. employment. The increase in China's exports, which accelerated after its admission to the World Trade Organization, is estimated to account for up to 25% of the drop in U.S. manufacturing employment (Pierce and Schott, 2012; Autor et al., 2013; Acemoglu et al., 2014).² Detailed studies at the worker level show that Chinese import competition has even affected long-term earning trajectories (Artuç et al., 2010; Autor et al., 2014). Whether the benefits of import competition outsize these costs is thus an open question, with important implications for policy making.

We contend we can learn about the implications of import competition by observing asset prices. If firms that face a larger displacement risk from import competition also have different risk premia than less-exposed firms, then asset prices inform us about how and how much do investors care about the risk of import competition. We find firms with greater exposure to that risk also command higher risk premia, suggesting the price of import competition risk is negative. States of the world where firms suffer from import competition are also states where consumption is dear for investors. Hence our result sheds light on the perceived benefits of openness to trade.

Within a standard model of trade flows we propose a mechanism through which investors may suffer from import competition rather than just benefiting from it through lower good prices. Investors' home bias limits their ability to do risk sharing efficiently and exposes them to import risk: as a foreign firm enters a domestic market and conquers market shares, investors' home bias prevents them from the natural risk sharing of a globally diversified portfolio.³

¹Authors' computations based on Census data and NBER CES data.

²Frictions on the labor market seem to have prevented a quick reallocation of the workforce, so that U.S. regions most exposed to Chinese imports experienced higher unemployment, lower wages, and lower participation rates.

³For evidence of home bias in U.S. investors' portfolio, see Coval and Moskowitz (1999); Ivković and Weisbenner (2005); Rauh (2006); Brown et al. (2009); Baik et al. (2010); Bernile et al. (2015).

We start by sorting industries with respect to their exposure to displacement risk. We hypothesize that firms are less likely to be displaced if the shipping costs incurred to replace their products with foreign ones are larger. We follow Bernard et al. (2006b) and exploit import data which allows us to compute the various costs associated to shipments, called Cost-Insurance-Freight as a percentage of the price paid by the importer (CIF). We document substantial cross-sectional variation and time-series persistence in CIF, consistent with the idea that this proxy captures structural and slow-moving barriers to import competition.

We then show that output and employment in high CIF industries are much less sensitive to changes in tariffs, which is consistent with the idea that large shipping costs partly insulate U.S. industries from import competition. We measure tariff changes as the annual change in tariff rates, defined as the ratio of collected duties to the total value of imports in any given industry and year. We first show that a drop in tariffs by 1 percentage point increases import penetration by 1 percentage point over the next five years, but only in sectors with low CIF. These sectors experience significant drops in employment, shipment, and value added growth by respectively 2.3%, 3.1% and 3.5% over the subsequent five years. At the firm level, stock returns drop by 4% in annualized terms following a drop by 1 percentage points in tariffs, but again only in low CIF sectors. The intuition for this result is straightforward: imports are less elastic to changes in tariffs in sectors with high CIF mark-ups. As a result, firms and employment are less displaced in these sectors.

Motivated by these results, we develop additional pricing predictions within the standard Melitz-Chaney model of trade. First we derive the elasticities of industry profit to the cost of bilateral trade. Then we characterize the effect of a change in that cost on domestic households' utility. Under our limited risk sharing assumption, we show their utility is subject to two competing effects: a positive price effect where the price of the final consumption index decreases as import competition intensifies; a negative income effect due to the decrease in households' wealth since the value of the domestic portfolio drops after an increase in import competition. The sum of both effects on final utility is ambiguous: the sign of the risk premium associated with the risk of import competition would determine if the price or the income effect dominates.

We compute monthly four factor alpha of five stock portfolios based on their industry CIF in the previous year. We find that the lowest CIF portfolio has an abnormal returns of 35 basis points, and that the hedge portfolio (high minus low CIF) generates abnormal returns of 57 basis points per month, or over 7% in annualized terms. Following the guidance of the model, we split the sample further into terciles of size, return on asset, and fixed costs intensity. We proxy for the intensity of fixed costs using two alternative proxies, namely the correlation of sales growth and cost growth in the past five to ten years, and the ratio of

sales, general, and administrative expenses (SGA) to sales. Consistent with the theoretical predictions, we find that the abnormal returns are concentrated among small firms with low return-on-assets and high fixed costs. Furthermore, we show that abnormal returns are found in non concentrated industries, another prediction of our theoretical framework. We also run Fama-McBeth regressions of monthly stock returns on the value of CIF (rather than quintiles of CIF), and we find similar effects. Taken together, the results indicate that stocks more exposed to import competition earn higher returns. This suggests that displacement risk covaries positively with the marginal utility of the representative investor.

We contribute to the literature, which starting with Melitz (2003) and Bernard et al. (2003), has taken into account firm heterogeneity to analyze the gains from trade.⁴ A common prediction of these models is that international trade elevates productivity through the contraction and exit of low-productivity firms and the expansion and entry into export markets of high-productivity firms. In this framework, globalization generates both winners and losers among firms within an industry, as better-performing firms expand into foreign markets, while worse-performing firms contract in the face of foreign competition. Consistent with this idea, Pavcnik (2002) finds that roughly two-thirds of the 19 percent increase in aggregate productivity following Chile's trade liberalization of the late 1970s and early 1980s is due to the relatively greater survival and growth of high-productivity plants. Bernard and Jensen (2004) find that almost half of all U.S. manufacturing productivity growth during 1983-1992 is explained by the reallocation of resources towards exporters. Trefler (2004) shows that 12 percent of the workers in low-productivity firms lost their jobs after the Canada-U.S. free trade agreement.

We also build on recent work that points out the displacement risk associated with imports. Bernard et al. (2006a) find that exposure to low-wage country imports is negatively associated with plant survival and employment growth, and Bernard et al. (2006b) find that the probability of plant death is higher in industries experiencing declining trade costs. Our results also relate to recent studies of the effect on the labor market of the acceleration of Chinese import penetration (Pierce and Schott, 2012; Autor et al., 2013; Acemoglu et al., 2014; Autor et al., 2014), or of trade shocks more generally (Artuç et al., 2010; Ebenstein et al., 2014). Our contribution is to show that displacement risk is reflected in the cost of capital, which suggests that the marginal utility of the representative investor covaries positively with this risk.

Finally, we add to a growing literature in finance that focuses on the implications of product market dynamics, including international trade for asset pricing. Recent contributions include Hou and Robinson (2006), Tian (2011), Loualiche (2013), Ready et al. (2013).

⁴For recent reviews, see Bernard et al. (2007), Melitz and Trefler (2012), Melitz and Redding (2014).

A common result in these papers is that the threat of entry tends to be priced in the cross-section of expected returns. In addition, a series of papers have used tariff cuts to instrument for import competition and have found that it affects firms capital budgeting decisions (Bloom et al., 2011; Fresard and Valta, 2014), and capital structure (Xu, 2012; Valta, 2012). Firms have also been found to suffer less from import competition if they have larger cash holdings (Fresard, 2010) and R&D expenses (Hombert and Matray, 2014). In relation to these papers, we show that the mere threat of import competition has an effect on firms through their higher cost of capital.

The remainder of the paper is organized as follows. Section 2 presents the effect of tariff cuts on industry outcomes conditional on shipping costs. In Section 3, we lay out the theoretical framework. Section 4 describes the relationship between shipping costs and risk premia, and Section 5 concludes.

2 Shipping costs, tariff changes and real outcomes

2.1 Measuring shipping costs

We start by sorting industries with respect to their exposure to displacement risk. We hypothesize that firms are less likely to be displaced if the shipping costs incurred to replace their products with imported ones are larger. We measure these costs using the actual shipping cost paid by importers. For this, we follow Bernard et al. (2006b) and measure ad valorem freight rate from underlying product-level U.S. import data compiled by Feenstra (1996), available from 1975 to 2005. We extend this data to 2012 by using U.S. import data available from the Census and obtained from Peter Schott’s website. Freight costs – our proxy for shipping costs – is the markup of the Cost-Insurance-Freight (CIF) value over the Free-on-Board value.

Building on prior work, we argue that CIF is a structural characteristic rooted in the nature of the output produced by any given industry.⁵ According to Hummels (2007), CIF depends on distance, quality, and weight-to-value ratio, which are persistent and vary substantially across industries.⁶

⁵The main limitation of CIF is that it does not take into account unobserved shipping costs – for instance information barriers and contract enforcement costs, holding costs for the goods in transit, inventory costs due to buffering the variability of delivery dates, or preparation costs associated with shipment size (Anderson and van Wincoop, 2004) . Unless these costs are correlated in systematic ways with CIF, they are likely to introduce noise in our measure of the sectoral exposure to displacement risk, which should generate an attenuation bias in our results. For recent contributions to the literature that adopts a structural approach to measure trade costs and estimate their effect on trade, see for instance Hummels and Skiba (2004), Das et al. (2007), or Irarrazabal et al. (2013).

⁶In Table 5, we find that CIF is highly correlated to the weight-to-value ratio, measured as the ratio of

To check whether CIF is indeed slow moving, we sort sectors by quintiles of CIF each year, and look at the transition across quintiles over time. We present this analysis in Table 1. The first panel highlights the transition from year $t - 1$ to year t , while the second panel shows the transition from year $t - 5$ to year t . For sectors in the top or bottom quintiles of CIF, the probability of being in the same quintile in the following period is around 86%. To document the substantial heterogeneity of CIF across industry in our sample, we compute the mean CIF for the top and bottom quintiles of CIF. As evidenced from Figure 1, the CIF mark-up on the Free-on-Board value is 13% in high CIF industries, and approximately 2% in low CIF industries.

2.2 Empirical strategy

Next we confirm CIF is a relevant proxy for the exposure to the displacement risk associated to import competition. For this, we use tariff changes as plausibly exogenous shocks to the attractiveness of imported goods relative to domestically produced goods. If CIF acts as a protection against displacement risk, then import penetration and other outcomes should be less responsive to tariff changes in high than in low CIF sectors.

The key identification threat is that tariff changes might be endogenous to industry outcomes. This might be the case for at least two reasons: (i) past performance could trigger tariff changes; and (ii) policymakers might change tariffs in anticipation of future investment opportunities.

Tariff changes might depend on past industry outcomes if for instance policymakers decide to decrease tariffs in industries that have done particularly well in the past, because they are unlikely to be harmed much by tariff cuts. Alternatively, policymakers could instead give up on industries that have been doing poorly and reduce tariff barriers in those. Fortunately, we can directly check in the data whether tariff changes are or not correlated with past industry penetration and output growth.

The second source of endogeneity comes from the potential correlation of tariff changes with future investment opportunities. For instance, policymakers might cut tariffs in industries that they expect will do great in the future. If anything, this should bias estimates against finding an negative effect of tariff changes on import penetration, output and employment. If, instead, policymakers cut tariffs when they expect industries to do poorly irrespective of import penetration, then this might be a concern. However, in order to fully explain our results, it should be the case that they cut tariffs when they expect poor outcomes in low CIF sectors, but not in high CIF ones. An important identifying assumption

kilograms shipped to the value of the shipment.

for our estimates is therefore that the way in which expectations of future sector outcomes motivate tariff changes is the same for high and low CIF industries.

Finally, for our identification to be valid, it needs to be the case that the exclusion restriction is satisfied, namely that tariff changes affect import penetration and sector outcomes only through their effect on imports. In particular, it should not be the case that tariff changes in the U.S. are matched abroad in a way that would also affect the exports of U.S. firms, and differentially so for high and low CIF sectors. We check and find that (i) tariff changes do not significantly affect the exports of domestic firms, and that (ii) there is no differences for high and low CIF sectors. In addition, if it was the case that tariff changes also affect exports, it would probably go against finding the results that we document here: following a bilateral tariff cut, low CIF sectors should be exposed to higher imports, but should also benefit from the ease of entry into foreign markets.

Subject to this identifying assumption, we measure tariffs in each manufacturing sector and year following Bernard et al. (2006b) and Fresard (2010), as the ratio of customs duties to the Free-on-Board value of imports. A key variable of interest, *Tariff change*, is defined as the difference in tariffs with respect to the previous year.⁷

One important concern with the construction of the tariff change variable is that variations in the composition of products or importers within industries could in theory induce variation in effective tariffs even if the statutory tariffs remain constant.⁸ This would be potentially a concern if in a given industry, consumers shift to lower tariff goods when they expect the domestic production to worsen. Again, in order to fully explain our results, it has to be the case that consumers only behave this way in low CIF industries, but not in high CIF ones. However, we try to go one step further to assuage this concern. We first build another variable called *Large tariff change* which is equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. This variable is probably more likely to capture abrupt statutory tariff changes triggered by policy decisions, rather than gradual effective tariff changes due to the evolving composition of the bundle of imported goods. Moreover, we check that we find similar results when we run our specifications using exclusively the tariff cuts induced by the Uruguay round of the World Trade Organization (WTO) in the late nineties. Finally, we also run a similar experiment where we consider variations in the Dollar-Yuan exchange rate instead of instead using changes in tariffs.

⁷Following Fresard (2010) we do not use the yearly changes between 1988 and 1989 when the coding imports changed, and set them equal to zero.

⁸Another limitation is that we do not observe variations in non-tariff barriers. However, unless these barriers are correlated in systematic ways with tariffs, they are likely to introduce noise in our estimation of the effect of tariff changes on sector outcomes, which should generate an attenuation bias in our results.

With these concerns in mind, we estimate the following panel regressions using all observations in the top and bottom CIF sectors quintiles where i indexes sectors and t indexes years:

$$Y_{i,t+1,t+6} = \alpha + \beta \cdot \text{High-CIF}_{i,t} \cdot \Delta\text{Tariff}_{i,t} + \gamma \cdot \text{High-CIF}_{i,t} + \kappa \cdot \Delta\text{Tariff}_{i,t} + \eta \cdot X_{i,t} + \theta_i + \delta_t + \epsilon_{i,t},$$

where $Y_{i,t+1,t+6}$ is the change between year $t+1$ and year $t+6$ in sector i in the variables of interest,⁹ including import penetration, log employment, log shipment and log value added. $\text{High-CIF}_{i,t}$ is a dummy equal to one if the sector's CIF lies in the top quintile of the distribution in year t , and zero if it lies in the bottom quintile. $\Delta\text{Tariff}_{i,t}$ is the change in tariff between year $t-1$ and t . $X_{i,t}$ is a vector of sector level time-varying characteristics including the level of tariffs, import penetration, log employment, log value added and log shipments. θ_i and δ_t are sector and time fixed effects, respectively. Finally, $\epsilon_{i,t}$ is an error term. Standards errors are corrected for clustering at the sector level. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Employment, shipments, value added, are obtained from the NBER CES files. The coefficient of interest are κ , which measures the effect of the tariff change on low CIF sectors, and β , which captures the differential effect of the tariff change on high CIF sectors.

2.3 Results

Table 2 presents the summary statistics for high and low CIF sectors respectively, where the former are those which CIF lies in the top quintile of the distribution in a given year, and the latter are those which CIF lies in the bottom one. As already noted, high CIF sectors have a mark-up of 13% above the Fee-on-Board price, more than six times as large as low CIF sectors. The level of tariffs is only slightly higher in high CIF industries. Unsurprisingly, both penetration and changes in penetration are lower in less exposed sectors. This validates our idea that CIF protects from import competition. Low CIF sectors are larger than high CIF ones, and experience a lower employment and output growth over the sample period.

In Table 3, we present the effect of tariff shocks on sector outcomes, conditional on the level of CIF. The results presented in Column 1 of the first panel show that a 1 percentage point decrease in tariffs leads to a 1 percentage point increase in import penetration over the next five years. However, tariffs have no effect on the import penetration of high CIF

⁹We choose a five year window following Bernard et al. (2006b).

sectors. This confirms that CIF act as a protection against import penetration. In Columns 2, 3, and 4, we consider the effect of tariff changes on employment, shipment and value added growth. We find that a 1 percentage point increase in tariffs leads to an increase by 2.1% in employment growth, by 3.1% in shipments growth, and by 3.5% in value added growth, but only in low CIF industries. Instead, high CIF sectors do not experience any significant change in these outcomes.

As mentioned above, a concern with our identification strategy is that tariff changes might be triggered by long-term trends. However, we show in Table B.1 in the Appendix that if we replace the tariff change from year $t - 1$ to t with the tariff change from year $t + 4$ to $t + 5$, it has no predictive power for the evolution of import penetration, employment, shipment or value added. Hence, tariff changes do not seem to be responding to trends in any systematic ways, for either high or low CIF sectors.

As already mentioned above, another concern for the identification relates to the measurement of tariff changes. If tariff changes capture differential endogenous recomposition of the import bundle in high and low CIF sectors, then our estimates might be biased. We attempt to address this concern in two ways. We first restrict the identification to large tariff changes only, which are more likely to be due to abrupt statutory tariff changes. We define a large tariff change as a variation at least as large as two times the median absolute change over the sample period. The result of this experiment is presented in the second panel of Table 3. Reassuringly, the estimates remain unaffected.

To make sure that the effect we are picking up are due to changes in statutory tariff changes, we then contrast CIF with a dummy taking the value of one in the years 1995 to 1998, when most of the tariff cuts associated to the Uruguay round of the World Trade Organization took place. Table B.2 in the Appendix lays out the results. Column 1 shows that in these three years, the average tariff change is -0.2% across sectors, and that there is no difference between high and low CIF sectors. However, as shown in column 2, the increase in import penetration following tariff cuts is much less pronounced in high CIF sectors (by 5.2 percentage points less). High CIF sectors are found to experience much better outcomes in the subsequent five years years: relative to low CIF sectors, employment, shipments and value added grow by between 5 and 7%.

In another robustness check, we consider the effect of appreciations of the Dollar against the Yuan, instead of using tariff changes. When the Dollar appreciates against the Yuan, this should make imports less costly and facilitate displacement of domestic firms, but only in highly exposed industry. Consistent with the results presented so far, we show in Table B.3 in the Appendix that increases in the Yuan–Dollar parity lead to lower growth in import penetration, and higher growth in employment, shipments and value added in industries

with high shipping costs, and therefore lower exposure.

Finally, we also ask whether the adverse effects that we document on industry dynamics translate into cash-flow losses, by looking at stock returns. If firms which experience a displacement shock due to import competition cannot instantaneously reallocate their inputs to other uses, they should experience negative returns when the shock materializes. This is exactly what we test in Table 4, where we run the same regressions than earlier at the firm level. The coefficient on the tariff change variable indicates that a 1 percentage point decrease in tariffs leads to a decrease by 0.34% in monthly stock returns in the year of the announcement, or nearly 4% in annualized terms. When we introduce stock fixed effects, the coefficient remains highly significant and the point estimates increases slightly. Importantly, these results are only found for low CIF sectors, that are highly exposed to import competition. We fail to find any significant effect of tariff changes on stock returns in high CIF industries. As placebo test, we replace tariff change from year $t - 1$ to t with the tariff change from year $t + 4$ to $t + 5$. Reassuringly, Table B.4 in the Appendix shows that this variable has virtually no predictive power for stock returns in year t .

3 Model

To build intuition about the role of the risk of import competition, we analyze the role of changes in trade costs within a standard model of trade flows. We follow the trade literature and develop our ideas in the workhorse model of (Chaney, 2008; Melitz, 2003). The model is static and we derive all our implications from comparative statics.

3.1 Setup

We start by setting up the model, which is solved in appendix (A). As in Chaney (2008), there are N countries that produce goods using labor as sole input. Each country has a labor force L_n , that determines the size of its economy. In each country consumers derive utility from the consumption of goods across $H + 1$ industries. Industry serves as a numeraire; there is a single good produced in industry 0, and it is freely tradable such that its price is unique across countries. In the H other industries multiple firms coexist and produce differentiated varieties of the same good. Households' utility of consuming the set $q_n^h(\cdot)$ of differentiated variety in industry h is summarized according to a constant elasticity of substitution (CES)

aggregator:

$$Q_n^h = \left[\int_{\Omega_n^h} q_n^h(\omega)^{\frac{\sigma_h-1}{\sigma_h}} d\omega \right]^{\frac{\sigma_h-1}{\sigma_h}},$$

where σ_h represents the industry specific elasticity of substitution across varieties, and Ω_n^h is the set of varieties available to households in industry h of country n . Finally the upper-tier utility \mathcal{U} over the $H + 1$ industries is of the Cobb-Douglas form:

$$\mathcal{U}_n = q_0^{\mu_0} \prod_{h=1}^H (Q_n^h)^{\mu_h},$$

where μ_h represents the expenditure shares of each industry, when we impose $\sum_{h \geq 0} \mu_h = 1$.

Supply Side — The homogenous good, in industry 0, is traded freely and serves as the numeraire in the global economy. Hence the relative productivity of each country for the good pins down the local wage rate w_n . For the other H industries, production is simple as firms operate a linear technology in labor. Within an industry firms differ by their productivity φ . Firms can produce so as to export into another country. We define a market as a triplet $\{j, i, h\}$ of firms from country j exporting into country i in industry h . Firms face two types of costs, variable *iceberg* costs, τ and fixed costs f that are both market specific. Thus the cost of producing q units of a good in market $\{j, i, h\}$ is:

$$c_{ji}^h(q; \varphi) = \frac{w_j}{\varphi} \tau_{ji}^h q + f_{ji}^h.$$

Iceberg costs are such that for each unit of the good produced only a fraction $1/\tau$ makes it to the importing country. The fixed costs are market specific as they represent the overhead of a firm forin a market.¹⁰

Within each industry firms operate in a monopolistically competitive environment: they take households' demand curve as given and set their prices accordingly. Given households' constant elasticity of substitution, σ_h , across varieties, firm prices are set at a constant markup over marginal cost:

$$p_{ji}^h(\varphi) = m_h w_j \tau_{ji}^h / \varphi,$$

where $m_h = \sigma_h / (\sigma_h - 1)$ is the markup in industry h .

Firm productivity is random; firms draw their productivity level φ upon entry into an

¹⁰We rule out triangular arbitrage by imposing $\tau_{ik} \leq \tau_{ij} \cdot \tau_{jk}$

industry from a Pareto distribution with tail parameter γ_h :¹¹ the probability of a draw below a given level φ , is:

$$\Pr\{\tilde{\varphi} < \varphi\} = G_h(\varphi) = 1 - \varphi^{-\gamma_h}.$$

Our framework is static. We do not allow for firm entry that could be endogenous to the industry structure or profits.¹² Hence we assume there is a fixed supply of entrants at the industry level; as in Chaney (2008) or Eaton and Kortum (2002) we assume the supply of entrants is proportional to the size of the domestic economy. Hence firms earn profits from their monopolistic position. We are interested in higher frequency movements where the supply of entrants is relatively inelastic. So movements in profits are largely due to entry and exit of existing firms into a market.

Equilibrium Quantities — Our main interest lies in the firms' profit functions and how they respond to changes in the competitive structure. Firm profits depend directly on the elasticity of substitution across goods in an industry and their idiosyncratic productivity φ . The building block is the local firm profit from operating in market $\{j, i, h\}$:

$$\pi_{ji}^h(\varphi) = \frac{\mu_h Y_i}{\sigma_h} \cdot \left[\frac{\sigma_h}{\sigma_h - 1} \frac{w_j \tau_{ji}^h / \varphi}{P_i^h} \right]^{1-\sigma_h} - f_{ji}^h.$$

where P_i^h is the price index of all varieties in industry h of country i . The equilibrium price index is simply $P_i^h = \kappa_1^h \cdot \theta_i^h \cdot Y_i^{\frac{1}{\gamma_h} - \frac{1}{\sigma_h - 1}}$. κ_1^h is a constant defined in appendix A.1. The coefficient θ_i^h represents an index of the remoteness of country i , it is expressed as a function of the weighted trade costs on market $\{k, i, h\}$, ϑ_{ki}^h as

$$\theta_i^{-\gamma_h} = \sum_k \vartheta_{ki}^h,$$

where $\vartheta_{ki}^h = w_k L_k (w_k \tau_{ki}^h)^{-\gamma_h} f_{ki}^{1 - \frac{\gamma_h}{\sigma_h - 1}}$

From the profit function we understand why firms get in and out of markets. If φ is too low a firm's profit cannot cover the fixed cost of operation in the market. Hence a firm's productivity level determines if they enter a market or not. We define the productivity cutoff for market $\{j, i, h\}$ as $\underline{\varphi}_{ji}^h = (\pi_{ji}^h)^{-1}(0)$. We detail the full expression of the productivity cutoff in the appendix. The cutoff productivity $\underline{\varphi}_{ji}^h$ is such that only firms with productivity above

¹¹The Pareto distribution assumption follows Chaney (2008); it reflects the actual distribution of firm sizes in the U.S.

¹²See Loualiche (2013) for a dynamic analysis in a closed economy.

it choose to enter the market. That cutoff represents a second margin of adjustment of trade flows to changes in trade costs: the extensive margin. If a market’s cutoff becomes larger because of an increase in trade costs than all supramarginal firms stop their operation on that market.

However the key quantity of interest for us is the average profit in an industry as it is what we observe empirically (see table 3). To get the average profit we integrate over all the productivity levels φ :

$$\pi_{ji}^h = \int \pi_{ji}^h(\varphi) dG_h(\varphi) = \frac{\mu_h}{\gamma_h} \cdot \frac{\sigma_h}{\sigma_h - 1} \cdot Y_i \cdot (w_j \tau_{ji}^h)^{-\gamma_h} (f_{ji}^h)^{1 - \frac{\gamma_h}{\sigma_h - 1}} (\theta_i^h)^{\gamma_h},$$

such that total aggregate profits is simply:

$$w_j L_j \pi_{ji}^h = \frac{\mu_h}{\gamma_h m_h} \cdot \frac{\vartheta_{ji}^h}{\sum_k \vartheta_{ki}^h} \cdot Y_i.$$

Profit is higher in larger export markets (large Y_i) and whenever both countries are “relatively” close to each other as summarised by ϑ_{ji}^h compared to the other distances.

3.2 Consequences of a Change in Trade Costs

A change in tariff — In section (2) we detailed empirically the consequences of a shock in tariffs for domestic firms. We reevaluate the results theoretically in the light of the Melitz-Chaney model. Then we explore which economic characteristics affect the elasticity of profits to a change in tariffs and more generally a change in the terms of trade on market $\{j, i, h\}$.

$$-\frac{\partial \log \pi_{ii}^h}{\partial \log \tau_{ji}^h} = -\gamma_h \cdot \alpha_{ji}^h,$$

where, $\alpha_{ji}^h = \frac{\vartheta_{ji}^h}{\sum_k \vartheta_{ki}^h}$

In industry h , the distance weighted share of country j for country i is α_{ji}^h . For example, if h is say the energy sector and country j is the largest world gas producer, then its contribution to industry h in country i will be large and α_{ji}^h will be closer to one. So the effect of a decrease in tariffs from country i to country j has adverse effects on the average firm’s profit in country i . The elasticity of average profits to tariffs is increasing in γ_h , the tail parameter of the firms’ productivity distribution: if γ_h is large, the industry is more homogeneous and a larger share of the output is concentrated among less productive firms. In that case the displacement from import competition is strongest. To understand the heterogeneous

effect of a decline in tariffs on firms of country i , we estimate the change of the productivity threshold for domestic production φ_{ii}^h . Movements in the productivity threshold correspond to displacement at the extensive margin, i.e. firms shutting down their operation in a specific market. We estimate the elasticity of the extensive margin to tariffs:

$$-\frac{\partial \log \varphi_{ii}^h}{\partial \log \tau_{ji}^h} = \alpha_{ji}^h. \quad (3.1)$$

Hence whenever tariffs decrease, the productivity threshold increases. The extent of this movement depends on the relative importance of country j for production of good h in country i , α_{ji}^h . Now a decrease in tariffs also affects the intensive margin, and even though firms above the productivity threshold stay in business, they lose market shares. The effects on profits at the individual firm level are:

$$-\frac{\partial \log \pi_{ii}^h(\varphi)}{\partial \log \tau_{ji}^h} = (\sigma_h - 1) \alpha_{ji}^h \cdot \left(1 + \frac{f_{ii}^h}{\pi_{ii}(\varphi)} \right). \quad (3.2)$$

The effects are strongest when the households' demand curve is elastic, that is whenever the elasticity of substitution σ_h is high. Moreover the elasticity is decreasing with profitability but increasing with the fixed costs at the industry level.

A change in import competition — More generally we are interested in the domestic response of a change in the terms of trade in market $\{j, i, h\}$. Our goal is to assess how a change in import competition affect the domestic incumbents. To quantify this margin, we derive the elasticity of both the extensive and intensive margin of domestic firms' operation to a decrease in the cost of labor in country j (or an increase in relative productivity in country j):

$$-\frac{\partial \log \varphi_{ii}^h}{\partial \log w_j} = \left(1 - \frac{1}{\gamma} \right) \cdot \alpha_{ji}^h. \quad (3.3)$$

$$-\frac{\partial \log \pi_{ii}^h(\varphi)}{\partial \log w_j} = (\sigma_h - 1) \left(1 - \frac{1}{\gamma} \right) \cdot \alpha_{ji}^h \cdot \left(1 + \frac{f_{ii}^h}{\pi_{ii}(\varphi)} \right). \quad (3.4)$$

In line with a decline in tariffs, domestic profits decrease after a shock to import competition. In our first empirical section we have established the role of shipping costs as moats: they protect incumbents from the displacement of foreign firms. From both elasticities (3.3) and (3.4)) we confirm our results and formulate a more general hypothesis we test in

Section (4).

Firms in industries with higher shipping costs (or other variable costs τ_{ji}^h) are shielded from import competition. Both elasticities decline with an increase in variable costs. The results stems from the role played by α_{ji}^h , the relative importance of country j for country's i consumption of goods in industry h . The elasticities are large whenever country j is a relative large exporter to country i . Whenever the level of shipping costs is high in an industry the role of country j declines and so does the impact of a shock of import competition from country j . Hence firms in industries with lower shipping costs are more exposed to the displacement risk of import competition than firms in high shipping costs industries. This is best summarized by the elasticity of relative importance of country j to variable costs:

$$-\frac{\partial \log \alpha_{ji}^h}{\partial \log \tau_{ji}^h} = \gamma_h \cdot \left(1 - \frac{\vartheta_{ji}^h}{\sum_k \vartheta_{ki}^h} \right) \quad (3.5)$$

Furthermore the elasticity of profits to import competition in equation 3.4 provides further empirical predictions not foreseen by our initial empirical analysis: firms with higher levels of fixed costs (f_{ii}^h) are more sensitive to displacement risk, their elasticity to import competition is greater than firms with low fixed costs; firms with low productivity are also more sensitive since either they cease to operate (extensive margin channel) through (equation 3.3) or their cash-flows decline through greater competition (equation 3.4).

Now we turn to the general equilibrium implications of the model. We have established import competition is a source of risk for domestic incumbents, especially in low variable trade costs industries. However to predict the price attached to that risk, we need to understand how and how much investors care about it.

3.3 Role of trade shocks for aggregate risk

In a perfect risk sharing economy, a decrease in trade costs is welfare improving. However the assumption of openness to trade as uniformly welfare improving has come under increasing scrutiny in the recent literature (see for example Autor et al. (2013)).

In this section, we propose a mechanism through which households might suffer from import competition, even though it improves their consumption basket. We assume households suffer from home bias when deciding on their stock portfolio investments: they do not invest in foreign firms. Under this assumption there is only limited risk sharing in the global economy. We show households are ambivalent about an increase in import competition: on the one hand it lowers the price of consumption good ($-\partial P_i^h / \partial w_j < 0$), what we refer to as the “price effect”. On the other hand, it displaces incumbent domestic firms by stealing

their market shares, hence it lowers the total wealth of domestic households ($-\partial Y_i/\partial w_j < 0$), what we refer to as the “income effect”.

To understand the trade-off faced by households, we estimate the change in domestic utility, \mathcal{U}_i , after an increase in import competition. We decompose the total effect on utility between a price effect (positive) and an income effect (negative):

$$-\frac{\partial \log \mathcal{U}_i}{\partial \log w_j} = \sum_h \mu_h \left(1 - \frac{1}{\gamma_h}\right) \alpha_{ji}^h \left[1 - \left(\sum_l \mu_l \left(1 - \frac{1}{\gamma_l}\right)\right) \cdot \frac{\frac{1}{m_h \gamma_h} \alpha_{ii}^h}{1 - \sum_l \frac{\mu_l}{m_l \gamma_l} \alpha_{ii}^l}\right]$$

The income effect dominates whenever the industries being displaced constitute a large part of country i economy, that is if α_{ii}^h is large enough. Furthermore the income effect is strongest whenever γ_h and σ_h are big. That is whenever displacement is severe at the intensive and at the extensive margin.

To summarize, within a standard Melitz-Chaney model of trade flows, we are able to formulate two main predictions about asset prices: first we confirm the results of section 2, that firms in industries with higher trade barriers are insulated from potential tariff shocks or any other shocks that would affect import competition. Second import competition affects domestic aggregate consumption. Hence firms with lower trade barriers have a higher exposure to the aggregate risk of import competition. The sign of the price of risk depends on the sign of the impact of import competition on the contemporaneous utility. If the income effect dominates (which is negative), then import competition has an adverse effect and the price of risk is negative. In that case investors will command higher risk premia for holding stocks in firms within industries with low trade barriers. The risk premia would be of the opposite sign were the price effect to dominate. In the subsequent section, we build on our theoretical framework to understand the sign of the risk of import competition.

4 CIF and the cost of capital

4.1 The Cross-Section of Industry Returns

In this section we explore the asset pricing predictions from the model that industry level heterogeneity in trade costs are associated with differences in risk premia. To test this hypothesis, we form stock equally-weighted portfolios based on the firms’ industries shipping costs (CIF) in the previous year. In Table (5) we present summary statistics for the returns on the five portfolios. We find that firms in industries with low shipping costs have average returns that are 5 percent higher (annually) than average returns in the high shipping costs

industry. The Sharpe ratio of the long-short portfolio (column 6) is 27 percent. Once we adjust the portfolio for risk using a three factor model, we find that the long-short portfolio alpha is 0.57 basis point (7.1 percent annually). In Table B.6 in the Appendix, we find similar results when we sort firms in quintiles based on average weight-to-value ratios rather than shipping costs. These results show that firms in high-shipping costs industries have lower returns than in low-shipping cost industries. This accredits our theory that firms when exposed to the risk of import competition earn higher risk premia.

In Table B.5, we present results for value-weighted returns. In this case the point estimate of the risk adjusted return for the long-short portfolio is lower. The discrepancy between the value and equally-weighted returns Table is due to an overweighting of larger firms. In our later more detailed analysis we find firms with higher productivity and size are less exposed to displacement risk such that it biases our estimates downwards.

Further we use the model's predictions to dissect the source of cross-sectional risk premia. Within industries already sorted by their level of shipping costs, we separate firms based on a measure of their productivity and a measure of the level of fixed costs in the industry. From equation (3.2) we have shown firms with the lowest productivity have the most to fear from import competition, be it because of displacement at the intensive or extensive margin. Moreover the elasticity of profits to import competition is also increasing with the level of fixed costs in the industry. Hence we should observe firms in high fixed costs industries or with low productivity should have the highest exposure to the import competition risk.

We proxy productivity using size and return-on-assets (ROA). For the intensity of fixed costs, we use two measures: the correlation of sales growth and cost growth in the past five to ten years, and the ratio of sales, general, and administrative expenses (SGA) to sales. We present results for our double-sorted portfolios in Table 7. We report the four factor alpha for each of the portfolios. For the size sorted portfolios, we find in the lowest size tercile, a portfolio that goes long high shipping costs and short low shipping costs has an alpha of -101 basis point monthly. This difference decreases to -28 basis point for the highest size tercile. Similarly across terciles of ROA, we find the long-short portfolio alpha is -73 basis point while it falls to -36 basis point for firms with high ROA. Hence firms that are larger or more productive have a higher exposure to the risk of import competition. Regarding the fixed costs results, we confirm the predictions from the model and find that firms in industries with the highest level of fixed costs have higher exposure to import competition: the four factor alpha on the long-short portfolio is systematically higher for within high-fixed industries.

We test the robustness of these results in various ways. We first use quintiles of firms' characteristics and find comparable results that we present in Table B.8. We run the same

analysis on value-weighted portfolios. We find in Table B.7 that high CIF firms have significantly lower abnormal returns in the bottom tertile of firm size and return-on-assets. We also run Fama-McBeth regressions of monthly stock returns on the value of CIF (rather than quintiles of CIF). The results of this analysis are presented in Table 8. A one percentage point increase in CIF leads to a drop by 0.04% in monthly expected returns. This is consistent with the estimates obtained in Table 5. The effect of CIF on expected returns is the largest for stocks that lie in the lowest tercile of size, return-on-assets, and in the highest tercile of fixed costs. Taken together, the results indicate that stocks more exposed to import competition earn higher returns. This suggests that displacement risk covaries positively with the marginal utility of the representative investor.

Another prediction of the model is that risk premia should depend on industry structure. From equation (3.4), we know that the effect on domestic firms' profits of a shock to import competition depends on γ , the parameter of the tail parameter of the distribution of firms' productivity. Intuitively, displacement risk will be lower in an industry where the firm sized distribution is highly skewed to the right. To check whether this is indeed the case, we split firms in each CIF quintile into those belonging to high versus low concentration industries, measured with the Herfindahl-Hirschman Index (HHI) at the four-digit SIC level. We present the results in Table 9. Consistent with the model's prediction, we find that abnormal returns are only found in non concentrated industries.

4.2 Pricing of the Risk of Import Competition

Finally the model, even static, suggests a linear pricing model of the form

$$m = a - b^{\text{MKT}} R^{\text{MKT}} - b^{\text{Import}} \Delta \text{Import}.$$

We estimate the price of risk of the import competition shock implied by our model.

In Table 10, we report the result of the second stage GMM estimates of b^{MKT} and b^{Import} . We use the pricing errors on our set of test assets as moment conditions. Industry portfolios are natural test assets as our economic characteristics are industry based rather than at the firm level. Furthermore, industry returns do not display a strong factor structure (Lewellen et al., 2010), hence they have a fair amount of heterogeneity. To measure the shock to import competition we use the returns on the CIF portfolio as a proxy: the portfolio that is long high-shipping costs industries and short low-shipping costs industries.

Our estimates of the price of risk are negative and significant. We find the price of risk is -0.61 when we estimate a two factor model. When we estimate a four-factor model, with the three Fama-French factors and the CIF portfolio, the price of risk is -0.369, inline with

the Sharpe ratio of the long-short portfolio of Table 5. Both of our estimates are strongly statistically significant. We conclude that import competition risk is priced in the cross-section of industry returns, and its price is economically significant. Import competition displaces firms' cashflows at times where consumption seems to be desirable.

5 Conclusion

The dramatic increase in import penetration is among the most important changes which affected the U.S. economy over the past decades. However, whether the benefits of import competition outsize these costs is an open question, with important implications for policy making. We contend that we can learn about the implications of import competition by observing its impact on asset prices. Our simple argument is that if the marginal utility of the representative investor goes up at times when import competition intensifies, then assets facing a larger displacement risk should command higher returns, and conversely. Unless the representative investor holds the global portfolio, measuring the excess returns on sectors highly exposed to displacement risk can tell us whether the covariance between this risk and marginal utility is positive or negative.

We investigate how the displacement risk associated with import competition is indeed reflected in the cost of capital. We sort U.S. industries on their exposure to import competition, based on shipping costs. We find that the output and employment in high exposure industries is more sensitive to tariff cuts than in low exposure industries, consistent with the idea that they face a higher risk of being displaced by import competition. Finally, we show that high exposure industries have a higher cost of capital. We can thus confirm that displacement risk of import competition is priced and covaries with the marginal utility of the representative agent.

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6 Figures

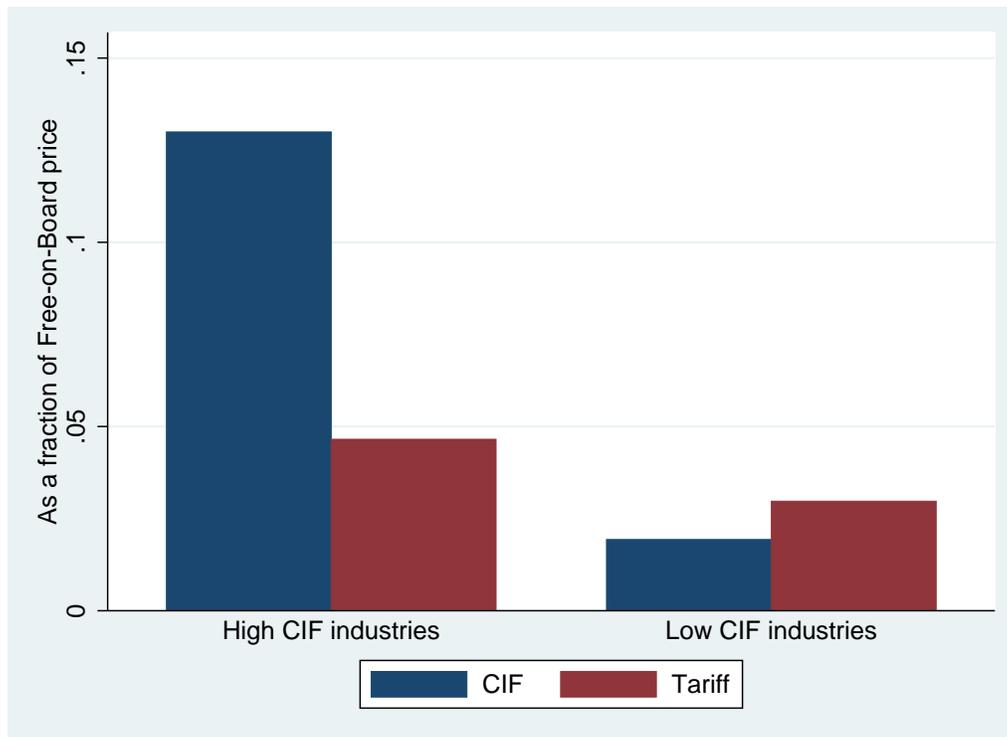


Figure 1
CIF and tariffs

This figure presents shipping costs (CIF) and tariffs for high and low CIF industries. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year.

7 Tables

Table 1
CIF persistence

This table presents the frequency of transition across shipping cost quintiles from year $t - 1$ to t and $t - 5$ to t in the sample. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports.

Transitions from year $t - 1$ to year t					
	Q1 (t)	Q2 (t)	Q3 (t)	Q4 (t)	Q5 (t)
Q1 (t-1)	85.0	12.2	1.5	0.6	0.8
Q2 (t-1)	12.2	72.5	13.4	1.5	0.4
Q3 (t-1)	1.3	13.4	67.1	16.5	1.7
Q4 (t-1)	0.6	1.4	16.3	70.6	11.1
Q5 (t-1)	0.7	0.7	1.5	11.0	86.1
Transitions from year $t - 5$ to year t					
	Q1 (t)	Q2 (t)	Q3 (t)	Q4 (t)	Q5 (t)
Q1 (t-5)	72.3	18.4	4.7	2.4	2.1
Q2 (t-5)	18.0	54.6	20.6	4.9	2.0
Q3 (t-5)	4.4	19.5	47.7	24.0	4.4
Q4 (t-5)	1.9	6.0	22.1	51.7	18.2
Q5 (t-5)	1.5	2.0	5.1	17.6	73.8

Table 2
Summary statistics

This table presents the summary statistics for the industry-year sample. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Tariff change is the difference in tariffs with respect to the previous year. Large tariff change is a variable equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. Employment, shipments, value added, are obtained from the NBER CES files. The sample period is from 1974 to 2006.

	Low CIF		High CIF	
	Mean	Std. dev.	Mean	Std. dev.
CIF	0.019	0.007	0.130	0.054
Tariff	0.031	0.035	0.047	0.053
Tariff change	-0.002	0.006	-0.002	0.010
Large tariff change	-0.002	0.006	-0.002	0.010
Penetration	0.211	0.209	0.122	0.173
Log employment	3.335	1.124	2.749	1.128
Log shipments	8.275	1.306	7.742	1.299
Log value added	7.592	1.356	6.887	1.227
Total factor productivity	1.073	0.789	0.997	0.126
$\Delta_{t,t+6}$ CIF	0.003	0.020	-0.017	0.048
$\Delta_{t,t+6}$ Tariff	-0.008	0.019	-0.009	0.027
$\Delta_{t,t+6}$ Penetration	0.057	0.111	0.029	0.079
$\Delta_{t,t+6}$ Log employment	-0.112	0.267	-0.077	0.213
$\Delta_{t,t+6}$ Log shipments	0.158	0.340	0.175	0.245
$\Delta_{t,t+6}$ Log value added	0.168	0.361	0.203	0.289
$\Delta_{t,t+6}$ Total factor productivity	0.162	1.561	0.040	0.137

Table 3
Effects of tariff changes on cash-flows at the sector level

This table presents the result of industry-year panel regressions assessing the effect of tariff cuts on various outcomes, conditional on the level of shipping costs. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Tariff change is the difference in tariffs with respect to the previous year. Large tariff change is a variable equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. All regressions include control for the industry level of tariffs, penetration, log employment, log value added and log shipments. Employment, shipments, value added, are obtained from the NBER CES files. Standard errors are clustered at the industry level and reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1974 to 2006.

	Delta (t+1, t+6)			
	Import penetration	Log employment	Log shipments	Log value added
	All tariff changes			
Tariff change x High CIF	1.3*** (0.5)	-3.1*** (1.0)	-4.2*** (1.5)	-4.7*** (1.7)
Tariff change	-1.1** (0.4)	2.3*** (0.8)	3.1** (1.4)	3.5** (1.5)
High CIF	0.0* (0.0)	0.1* (0.0)	0.1** (0.0)	0.1** (0.1)
Controls	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Sector FE	Yes	Yes	Yes	Yes
Observations	3455	3455	3455	3455
R ²	0.306	0.513	0.493	0.432
	Large tariff changes			
Large tariff change x High CIF	1.2** (0.5)	-3.1*** (1.0)	-4.3*** (1.5)	-4.6*** (1.7)
Large tariff change	-1.0** (0.4)	2.3*** (0.9)	3.2** (1.4)	3.5** (1.5)
High CIF	0.0* (0.0)	0.1* (0.0)	0.1** (0.0)	0.1** (0.1)
Controls	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Sector FE	Yes	Yes	Yes	Yes
Observations	3455	3455	3455	3455
R ²	0.305	0.513	0.494	0.432

Table 4
Effects of tariff changes on stock returns at the firm level

This table presents the result of firm-level regressions assessing the effect of tariff cuts on the average monthly return in any given year, conditional on the level of shipping costs. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Tariff change is the difference in tariffs with respect to the previous year. Large tariff change is a variable equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. All regressions include control for the industry level of tariffs, penetration, log employment, log value added and log shipments. Employment, shipments, value added, are obtained from the NBER CES files. Standard errors are clustered at the industry level and reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1974 to 2006.

Dependent variable:	Average monthly return (t-1)	
	All tariff changes	
Tariff change x High CIF	-0.27* (0.16)	-0.16 (0.12)
Tariff change	0.34** (0.14)	0.26*** (0.09)
High CIF	-0.01*** (0.00)	0.01 (0.01)
Controls	Yes	Yes
Year FE	Yes	Yes
Firm FE	No	Yes
Observations	16645	16645
R^2	0.114	0.296
	Large tariff changes	
Large tariff change x High CIF	-0.35** (0.16)	-0.20 (0.13)
Large tariff change	0.40*** (0.13)	0.33*** (0.10)
High CIF	-0.01*** (0.00)	0.01 (0.01)
Controls	Yes	Yes
Year FE	Yes	Yes
Firm FE	No	Yes
Observations	16645	16645
R^2	0.114	0.296

Table 5
CIF Portfolios - Summary statistics

The table reports summary statistics for five portfolios of industries sorted on CIF. We report average shipping cost, portfolios' market shares computed using the sum of the market capitalization across all stocks that belong to the same portfolio, and equally-weighted book-to-market of each portfolio. We also report (annualized) mean excess returns over the risk-free rate (μ), volatilities (σ) and Sharpe ratios ($\mu/\sqrt{12}\sigma$). The sample period is from 1974 to 2013.

	CIF Portfolios				
	Low	2	3	4	High
Average Shipping Cost (CIF)	0.011	0.019	0.025	0.036	0.080
Weight-to-value ratio	0.009	0.023	0.042	0.118	1.205
B/M	0.613	0.725	0.762	0.927	1.078
Market share	0.328	0.147	0.133	0.142	0.251

	CIF Portfolios					
Portfolio Moments	Low	2	3	4	High	Hedge
Mean excess return (μ)	0.170	0.136	0.112	0.131	0.111	-0.051
Volatility (σ)	0.081	0.075	0.068	0.063	0.057	0.054
Sharpe ratio	0.604	0.524	0.472	0.600	0.569	-0.272

Table 6
CIF portfolios - Equally-weighted Returns

This table presents the monthly excess returns (α) over a four factor model of portfolios constructed based on the shipping costs (CIF) in their industry. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. In any given month, stocks are sorted into five portfolios based on their industry CIF in the previous year. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. Standard Errors are estimated using Newey-West with 12 lags. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. The sample period is from 1974 to 2013.

	CIF Portfolios					
	Low	2	3	4	High	Hedge
α	0.355*	0.060	-0.103	-0.064	-0.217*	-0.572**
	(0.208)	(0.160)	(0.128)	(0.109)	(0.115)	(0.279)
β^{MKT}	1.080***	1.043***	1.012***	1.087***	1.039***	-0.041
	(0.053)	(0.038)	(0.034)	(0.036)	(0.034)	(0.074)
β^{SMB}	-0.395***	-0.186***	-0.047	0.288***	0.560***	0.955***
	(0.055)	(0.061)	(0.065)	(0.070)	(0.077)	(0.110)
β^{HML}	1.228***	1.221***	1.066***	0.871***	0.675***	-0.553***
	(0.093)	(0.061)	(0.061)	(0.078)	(0.098)	(0.176)

Table 7
Equally-weighted CIF portfolios, conditional on cross-sectional characteristics

This table presents the equally-weighted monthly excess returns (Alpha) over a four factor model of portfolios constructed based on the shipping costs (CIF) in their industry. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. In any given month, stocks are sorted into five portfolios based on their industry CIF in the previous year. Firms are then sorted in terciles based on their market capitalization (SIZE), return on assets (ROA), as well as two measures of fixed costs (FIXED COSTS), namely the correlation of sales growth and cost growth in the past five to ten years, and the ratio of sales, general, and administrative expenses (SGA) to sales. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. Standard errors are estimated using Newey-West with 12 lags. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. The sample period is from 1974 to 2013.

	CIF Portfolios					
	Low	2	3	4	High	Hedge
	Size terciles					
T1	0.68** (0.29)	0.32 (0.21)	0.11 (0.21)	0.05 (0.18)	-0.33* (0.19)	-1.01*** (0.33)
T2	0.15 (0.25)	-0.09 (0.21)	-0.25 (0.16)	-0.16 (0.12)	-0.27** (0.12)	-0.42 (0.33)
T3	0.23 (0.20)	-0.05 (0.14)	-0.16 (0.12)	-0.08 (0.11)	-0.05 (0.11)	-0.28 (0.26)
	ROA terciles					
T1	0.32 (0.35)	-0.16 (0.23)	-0.33 (0.21)	-0.08 (0.17)	-0.41** (0.17)	-0.73* (0.43)
T2	0.48** (0.22)	0.36** (0.18)	0.12 (0.13)	-0.09 (0.12)	-0.08 (0.11)	-0.56* (0.29)
T3	0.41** (0.16)	0.28* (0.15)	0.13 (0.14)	0.04 (0.12)	0.05 (0.12)	-0.36 (0.23)
	Fixed costs terciles (1) - corr(sales growth, cost growth)					
T1	0.28 (0.21)	0.32** (0.16)	0.24* (0.13)	0.02 (0.12)	-0.10 (0.12)	-0.38 (0.28)
T2	0.60*** (0.22)	0.22 (0.17)	-0.03 (0.15)	0.01 (0.12)	-0.04 (0.12)	-0.64** (0.30)
T3	0.36 (0.24)	-0.07 (0.20)	-0.29* (0.17)	-0.15 (0.15)	-0.28** (0.14)	-0.65** (0.33)
	Fixed costs terciles (2) - SGA over sales					
T1	0.28 (0.20)	0.22 (0.15)	-0.10 (0.13)	-0.17 (0.15)	-0.18 (0.15)	-0.46* (0.27)
T2	0.62*** (0.23)	0.29* (0.15)	0.08 (0.15)	-0.09 (0.13)	-0.11 (0.12)	-0.73** (0.31)
T3	0.69*** (0.20)	0.01 (0.21)	0.05 (0.19)	0.17 (0.14)	-0.14 (0.13)	-0.83*** (0.24)

Table 8
Fama-MacBeth Return Regressions

This table reports the Fama-MacBeth coefficients from monthly cross-sectional regressions of individual stock returns on CIF and controls. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. Firms are then sorted in terciles based on their size, return on assets (ROA), as well as two measures of fixed costs, namely the correlation of sales growth and cost growth in the past five to ten years (Corr), and the ratio of sales, general, and administrative expenses (SGA) to sales. Size is the logarithm of last month stock market capitalization. Turnover is the logarithm of last month volume scaled by shares outstanding. All independent variables are winsorized at the 1st and 99th percentiles. RET(-1) is last month stock return. RET(-2,-12) is the cumulative stock return over the 11 months ending at the beginning of the previous month. (Simple) standard errors are reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1974 to 2013.

		Monthly stock returns							
		Size		ROA		Fixed costs			
		Low	High	Low	High	Low	High	Low	High
CIF	-0.04** (0.02)	-0.11*** (0.03)	-0.01 (0.02)	-0.11*** (0.03)	-0.01 (0.02)	-0.03* (0.02)	-0.06* (0.03)	-0.03 (0.02)	-0.05** (0.02)
Size	-0.00* (0.00)	-0.00*** (0.00)	-0.00 (0.00)	-0.00* (0.00)	-0.00 (0.00)	-0.00** (0.00)	-0.00 (0.00)	-0.00** (0.00)	-0.00 (0.00)
Turnover	0.00*** (0.00)	0.01*** (0.00)	0.00 (0.00)	0.00*** (0.00)	0.00 (0.00)	0.00*** (0.00)	0.00** (0.00)	0.00*** (0.00)	0.00*** (0.00)
Ln(B/M)	0.00*** (0.00)	0.01*** (0.00)	0.00*** (0.00)	0.01*** (0.00)	0.00*** (0.00)	0.00*** (0.00)	0.00*** (0.00)	0.00*** (0.00)	0.01*** (0.00)
Sales margin	0.00 (0.00)	0.01** (0.01)	-0.00 (0.00)	0.01* (0.01)	-0.00 (0.01)	0.00 (0.00)	0.00 (0.00)	0.00 (0.01)	0.00 (0.00)
RET(-1)	-0.05*** (0.00)	-0.08*** (0.01)	-0.03*** (0.01)	-0.06*** (0.01)	-0.04*** (0.01)	-0.06*** (0.01)	-0.04*** (0.01)	-0.04*** (0.01)	-0.05*** (0.01)
Constant	0.03*** (0.01)	0.07*** (0.01)	0.03*** (0.01)	0.03*** (0.01)	0.03*** (0.01)	0.03*** (0.01)	0.02** (0.01)	0.03*** (0.01)	0.02*** (0.01)
Observations	429682	122352	167548	127592	152470	147305	136135	137699	126459
R ²	0.051	0.065	0.084	0.065	0.070	0.072	0.064	0.069	0.068

Table 9
Equally-weighted CIF portfolios, conditional on industry concentration

This table presents the equally-weighted monthly excess returns (Alpha) over a four factor model of portfolios constructed based on the shipping costs (CIF) in their industry. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. In any given month, stocks are sorted into five portfolios based on their industry CIF in the previous year. Firms are then sorted based on industry concentration, measured with the HerfindahlHirschman Index of their four-digit industry. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. Standard errors are estimated using Newey-West with 12 lags. The sample period is from 1974 to 2013.

		CIF					
		Low	2	3	4	High	Hedge
		HHI (4-digit SIC)					
Low	0.43** (0.21)	-0.12 (0.18)	-0.16 (0.14)	-0.02 (0.12)	-0.26* (0.14)	-0.70** (0.29)	
High	0.21 (0.25)	0.28 (0.21)	-0.03 (0.15)	-0.11 (0.13)	-0.14 (0.12)	-0.35 (0.31)	

Table 10
GMM Estimate of a Linear Factor Model

Factor Price	(CAPM)	(CAPM + CIF)	(FF + CIF)
R_{MKT}^e	1.08 (0.126)	1.23 (0.138)	0.826 (0.15)
$\Delta\{R^e \text{CIF}\}$ eq. weighted		-0.61 (0.165)	-0.369 (0.118)

Standard Errors are estimated using Newey-West with 12 lags.

Appendix

A Model - Derivation

A.1 Solution

The price index is given by summing over all the prices for the varieties produced in country j , industry h :

$$(P_j^h)^{1-\sigma_h} = \sum_{k=1}^N w_k L_k \int_{\underline{\varphi}_{ki}^h}^{\infty} \left(\frac{\sigma_h}{\sigma_h - 1} \frac{w_k \tau_{kj}^h}{\varphi} \right)^{1-\sigma_h} dG_h(\varphi)$$

The remoteness index is given by:

$$(\theta_i^h)^{-\gamma_h} = \sum_{k=1}^N w_k L_k \cdot (w_k \tau_{ki}^h)^{-\gamma_h} (f_{ki}^h)^{1-\frac{\gamma_h}{\sigma_h-1}}$$

The price index solved for gives:

$$P_i^h = \left(\frac{\sigma_h}{\mu_h} \right)^{\frac{1}{\sigma_h-1} - \frac{1}{\gamma_h}} \left(\frac{\gamma_h}{\gamma_h - (\sigma_h - 1)} \right)^{-\frac{1}{\gamma_h}} \frac{\sigma_h}{\sigma_h - 1} \cdot \theta_i^h Y_i^{\frac{1}{\gamma_h} - \frac{1}{\sigma_h-1}}$$

The productivity cutoff is:

$$\underline{\varphi}_{ji}^h = \left(\frac{\sigma_h}{\mu_h} \right)^{\frac{1}{\sigma_h-1}} \frac{\sigma_h}{\sigma_h - 1} \cdot \left(\frac{f_{ji}^h}{Y_i} \right)^{\frac{1}{\sigma_h-1}} \cdot \frac{w_j \tau_{ji}^h}{P_i^h}$$

A.2 Notations

Hereafter we define some of the constants we use in the derivation of the model:

$$\kappa_1^h = \frac{\sigma_h}{\sigma_h - 1} \cdot \left(\frac{\gamma_h}{\gamma_h - (\sigma_h - 1)} \right)^{-\frac{1}{\gamma_h}} \cdot \left(\frac{\sigma_h}{\mu_h} \right)^{\frac{1}{\sigma_h-1} - \frac{1}{\gamma_h}}.$$

B Robustness tables

Table B.1
Effects of tariff changes on cash-flows at the sector level, placebo

This table presents the result of industry-year panel regressions assessing the effect of tariff cuts on various outcomes, conditional on the level of shipping costs. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Tariff change is the difference in tariffs with respect to the previous year. Large tariff change is a variable equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. All regressions include control for the industry level of tariffs, penetration, log employment, log value added and log shipments. Employment, shipments, value added, are obtained from the NBER CES files. Standard errors are clustered at the industry level and reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1974 to 2006.

	Delta (t+1, t+6)			
	Import penetration	Log employment	Log shipments	Log value added
	All tariff changes			
Tariff change (t+5) x High CIF	0.1 (0.6)	2.3 (1.5)	1.5 (2.2)	2.7 (2.2)
Tariff change (t+5)	0.0 (0.5)	-0.9 (1.3)	-1.2 (1.9)	-1.4 (1.9)
High CIF	0.0 (0.0)	0.1** (0.0)	0.1*** (0.0)	0.1** (0.1)
Controls	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Sector FE	Yes	Yes	Yes	Yes
Observations	3326	3326	3326	3326
R^2	0.312	0.541	0.531	0.457
	Large tariff changes			
Large tariff change (t+5) x High CIF	0.2 (0.6)	2.2 (1.5)	1.0 (2.2)	2.3 (2.2)
Large tariff change (t+5)	-0.1 (0.5)	-0.7 (1.3)	-0.6 (1.9)	-0.8 (1.9)
High CIF	0.0 (0.0)	0.1** (0.0)	0.1*** (0.0)	0.1** (0.1)
Controls	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Sector FE	Yes	Yes	Yes	Yes
Observations	3326	3326	3326	3326
R^2	0.312	0.541	0.531	0.457

Table B.2
Effects of WTO-induced tariff changes on cash-flows at the sector level

This table presents the result of industry-year panel regressions assessing the effect of tariff cuts on various outcomes, conditional on the level of shipping costs. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Tariff change is the difference in tariffs with respect to the previous year. Large tariff change is a variable equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. All regressions include control for the industry level of tariffs, penetration, log employment, log value added and log shipments. Employment, shipments, value added, are obtained from the NBER CES files. WTO is a dummy equal to one in years 1995 to 1998, when most of the tariff cuts associated with the Uruguay Round were passed. Standard errors are clustered at the industry level and reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1974 to 2006.

	Delta (t+1, t+6)				
	Tariff change	Import penetration	Log employment	Log shipments	Log value added
WTO x High CIF	0.001 (0.001)	-0.051*** (0.013)	0.079*** (0.027)	0.058* (0.034)	0.054 (0.039)
WTO	-0.002*** (0.000)				
High CIF	-0.001 (0.001)	0.035* (0.018)	0.067* (0.038)	0.095** (0.043)	0.111** (0.052)
Controls	No	Yes	Yes	Yes	Yes
Year FE	No	Yes	Yes	Yes	Yes
Sector FE	Yes	Yes	Yes	Yes	Yes
Observations	3455	3610	3610	3610	3610
R^2	0.128	0.306	0.513	0.503	0.436

Table B.3
Effects of appreciation of the Dollar against the Yuan on cash-flows at the sector level

This table presents the result of industry-year panel regressions assessing the effect of changes in the Dollar–Yuan exchange rate, conditional on the level of shipping costs. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. All regressions include control for the industry level of tariffs, penetration, log employment, log value added and log shipments. Employment, shipments, value added, are obtained from the NBER CES files. Standard errors are clustered at the industry level and reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1984 to 2006.

	Delta (t+1, t+6)			
	Import penetration	Log employment	Log shipments	Log value added
Dollar appreciation against the Yuan (t-1,t) x High CIF	-0.087** (0.034)	0.163** (0.072)	0.144* (0.076)	0.165** (0.083)
High CIF	0.008 (0.017)	0.065 (0.055)	0.129** (0.064)	0.157* (0.090)
Controls	Yes	Yes	Yes	Yes
Year FE	Yes	Yes	Yes	Yes
Sector FE	Yes	Yes	Yes	Yes
Observations	2245	2245	2245	2245
R^2	0.390	0.551	0.515	0.459

Table B.4
Effects of tariff changes on stock returns at the firm level, placebo

This table presents the result of firm-level regressions assessing the effect of tariff cuts on the average monthly return in any given year, conditional on the level of shipping costs. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. High (low) CIF industries are those in the top (bottom) quintile of the distribution of CIF in any given year. Tariffs are measured at the industry-year level as the ratio of customs duties to the Free-on-Board value of imports. Import penetration is measured at the industry-year level as the ratio of the Free-on-Board value of imports and the sum of total shipments and imports minus exports. Tariff change is the difference in tariffs with respect to the previous year. Large tariff change is a variable equal to the tariff change if it is larger than twice the median absolute tariff change in the sample, and zero otherwise. All regressions include control for the industry level of tariffs, penetration, log employment, log value added and log shipments. Employment, shipments, value added, are obtained from the NBER CES files. Standard errors are clustered at the industry level and reported in parentheses. *, ** and *** means statistically different from zero at 10%, 5% and 1% level of significance. The sample period is from 1974 to 2006.

Dependent variable:	Average monthly return (t-1)	
	All tariff changes at t+5	
Tariff change (t+5) x High CIF	-0.07 (0.18)	0.12 (0.24)
Tariff change (t+5)	-0.11 (0.14)	-0.27 (0.19)
High CIF	-0.01** (0.00)	0.01** (0.01)
Controls	Yes	Yes
Year FE	Yes	Yes
Firm FE	No	Yes
Observations	14335	14335
R^2	0.113	0.343
	Large tariff changes at t+5	
Large tariff change (t+5) x High CIF	0.00 (0.17)	0.15 (0.23)
Large tariff change (t+5)	-0.13 (0.14)	-0.27 (0.19)
High CIF	-0.01** (0.00)	0.01** (0.01)
Controls	Yes	Yes
Year FE	Yes	Yes
Firm FE	No	Yes
Observations	14335	14335
R^2	0.113	0.343

Table B.5
CIF portfolios - Value-weighted Returns

This table presents the monthly excess returns (Alpha) over a four factor model of portfolios constructed based on the shipping costs (CIF) in their industry. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. In any given month, stocks are sorted into five portfolios based on their industry CIF in the previous year. Monthly portfolio returns are value-weighted. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. Standard Errors are estimated using Newey-West with 12 lags. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. The sample period is from 1974 to 2013.

	CIF Portfolios					
	Low	2	3	4	High	Hedge
α	0.249*	0.024	-0.066	-0.184	0.117	-0.132
	(0.135)	(0.142)	(0.147)	(0.116)	(0.113)	(0.188)
β^{MKT}	0.933***	1.130***	1.069***	1.059***	0.893***	-0.039
	(0.050)	(0.037)	(0.043)	(0.027)	(0.039)	(0.060)
β^{SMB}	-0.369***	-0.332***	-0.302***	0.113*	0.240**	0.609***
	(0.053)	(0.068)	(0.096)	(0.064)	(0.097)	(0.103)
β^{HML}	-0.088*	0.214***	0.133**	0.143***	-0.046	0.042
	(0.053)	(0.070)	(0.054)	(0.043)	(0.073)	(0.103)

Table B.6
CIF portfolios - Weight-to-value ratio

This table presents the monthly excess returns (Alpha) over a four factor model of portfolios constructed based on the weight-to-value ratios in their industry. In any given month, stocks are sorted into five portfolios based on their industry weight-to-value ratio in the previous year. Monthly portfolio returns are value-weighted. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. Standard Errors are estimated using Newey-West with 12 lags. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. The sample period is from 1990 to 2013.

	Weight-to-value Portfolios					
	Low	2	3	4	High	Hedge
α	0.505	0.302	0.026	-0.164	-0.265	-0.771*
	(0.311)	(0.244)	(0.182)	(0.152)	(0.168)	(0.400)
β^{MKT}	1.142***	1.058***	1.076***	1.046***	1.065***	-0.077
	(0.078)	(0.055)	(0.062)	(0.042)	(0.054)	(0.096)
β^{SMB}	-0.457***	-0.317***	-0.168**	0.323***	0.693***	1.149***
	(0.065)	(0.069)	(0.084)	(0.088)	(0.075)	(0.102)
β^{HML}	1.240***	1.213***	1.052***	0.740***	0.605***	-0.635***
	(0.126)	(0.079)	(0.078)	(0.111)	(0.076)	(0.179)

Table B.7
Value-weighted CIF portfolios, conditional on cross-sectional characteristics

This table presents the value-weighted monthly excess returns (Alpha) over a four factor model of portfolios constructed based on the shipping costs (CIF) in their industry. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. In any given month, stocks are sorted into five portfolios based on their industry CIF in the previous year. Firms are then sorted in terciles based on their market capitalization (SIZE), return on assets (ROA), as well as two measures of fixed costs (FIXED COSTS), namely the correlation of sales growth and cost growth in the past five to ten years, and the ratio of sales, general, and administrative expenses (SGA) to sales. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. Standard Errors are estimated using Newey-West with 12 lags. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. The sample period is from 1974 to 2013.

	CIF Portfolios					
	Low	2	3	4	High	Hedge
Size terciles						
T1	0.43 (0.28)	-0.05 (0.19)	-0.06 (0.21)	-0.19 (0.18)	-0.57*** (0.19)	-1.00*** (0.34)
T2	0.10 (0.24)	-0.14 (0.21)	-0.26 (0.17)	-0.20 (0.13)	-0.25** (0.12)	-0.35 (0.32)
T3	0.26* (0.14)	0.04 (0.15)	-0.05 (0.15)	-0.18 (0.12)	0.13 (0.12)	-0.12 (0.19)
ROA terciles						
T1	0.12 (0.37)	-0.44 (0.27)	-0.29 (0.27)	-0.51** (0.24)	-0.23 (0.15)	-0.35 (0.41)
T2	0.13 (0.23)	-0.10 (0.19)	-0.11 (0.18)	-0.30* (0.16)	0.10 (0.14)	-0.03 (0.31)
T3	0.27* (0.15)	0.22 (0.18)	0.05 (0.17)	-0.03 (0.14)	0.23 (0.15)	-0.04 (0.22)
Fixed costs terciles (1) - corr(sales growth, cost growth)						
T1	0.28* (0.15)	-0.07 (0.11)	-0.01 (0.14)	-0.20* (0.11)	-0.05 (0.13)	-0.34* (0.18)
T2	0.13 (0.19)	0.22 (0.21)	-0.00 (0.24)	-0.31 (0.19)	0.25* (0.15)	0.12 (0.24)
T3	0.46* (0.26)	0.04 (0.21)	-0.30 (0.22)	0.02 (0.21)	0.10 (0.16)	-0.36 (0.35)
Fixed costs terciles (2) - SGA over sales						
T1	0.25 (0.20)	0.02 (0.15)	-0.06 (0.19)	-0.27 (0.18)	0.14 (0.15)	-0.11 (0.26)
T2	0.17 (0.17)	0.20 (0.19)	-0.12 (0.18)	0.02 (0.14)	0.04 (0.14)	-0.13 (0.25)
T3	0.57*** (0.22)	0.08 (0.20)	0.07 (0.22)	-0.07 (0.17)	0.02 (0.16)	-0.56** (0.23)

Table B.8
Equally-weighted trade cost portfolios, conditional on cross-sectional characteristics

This table presents the equally-weighted monthly excess returns (Alpha) over a four factor model of portfolios constructed based on the shipping costs (CIF) in their industry. CIF are measured at the industry-year level as the % difference of the Cost-Insurance-Freight value with the Free-on-Board value of imports. In any given month, stocks are sorted into five portfolios based on their industry CIF in the previous year. Firms are then sorted in quintiles based on their market capitalization (SIZE), return on assets (ROA), as well as two measures of fixed costs (FIXED COSTS), namely the correlation of sales growth and cost growth in the past five to ten years, and the ratio of sales, general, and administrative expenses (SGA) to sales. We regress a given portfolio's return in excess of the risk free rate on the market portfolio minus the risk-free rate, the size factor (small minus big), and the value factor (high minus low), all obtained from Kenneth French's website. Standard Errors are estimated using Newey-West with 12 lags. ***, **, and * indicate significance at the 1, 5, and 10% level, respectively. The sample period is from 1974 to 2013.

	CIF Portfolios					
	Low	2	3	4	High	Hedge
	Size quintiles					
Q1	0.93*** (0.34)	0.68** (0.27)	0.30 (0.23)	0.26 (0.21)	-0.15 (0.20)	-1.08*** (0.36)
Q2	0.34 (0.28)	-0.10 (0.20)	-0.16 (0.19)	-0.21 (0.16)	-0.52*** (0.18)	-0.86** (0.36)
Q3	0.13 (0.28)	-0.13 (0.22)	-0.28 (0.18)	-0.13 (0.14)	-0.27** (0.13)	-0.39 (0.36)
Q4	-0.00 (0.23)	-0.17 (0.21)	-0.27 (0.18)	-0.11 (0.13)	-0.04 (0.14)	-0.04 (0.32)
Q5	0.38** (0.19)	0.02 (0.13)	-0.12 (0.11)	-0.13 (0.12)	-0.10 (0.11)	-0.48** (0.24)
	ROA quintiles					
Q1	0.18 (0.37)	-0.45* (0.26)	-0.62*** (0.24)	-0.20 (0.18)	-0.61*** (0.20)	-0.79* (0.45)
Q2	0.57* (0.31)	0.31 (0.25)	0.17 (0.18)	0.03 (0.20)	-0.11 (0.15)	-0.69* (0.39)
Q3	0.43* (0.23)	0.45** (0.19)	0.08 (0.14)	-0.11 (0.12)	-0.17 (0.12)	-0.60* (0.31)
Q4	0.40** (0.20)	0.19 (0.16)	0.12 (0.14)	0.06 (0.12)	0.12 (0.12)	-0.28 (0.27)
Q5	0.43*** (0.16)	0.29* (0.16)	0.11 (0.15)	0.01 (0.13)	0.04 (0.12)	-0.39* (0.22)
	Fixed costs quintiles (1) - corr(sales growth, cost growth)					
Q1	0.19 (0.29)	0.25 (0.16)	0.26* (0.15)	0.00 (0.14)	-0.08 (0.14)	-0.27 (0.34)
Q2	0.45*** (0.16)	0.20 (0.16)	0.13 (0.13)	0.07 (0.14)	-0.13 (0.12)	-0.58** (0.24)
Q3	0.65*** (0.23)	0.31* (0.18)	-0.09 (0.17)	0.00 (0.13)	-0.01 (0.14)	-0.66** (0.30)
Q4	0.57* (0.30)	0.25 (0.22)	-0.11 (0.18)	-0.07 (0.16)	-0.13 (0.14)	-0.70* (0.39)
Q5	0.22 (0.24)	-0.23 (0.22)	-0.32 (0.20)	-0.20 (0.17)	-0.36** (0.16)	-0.58* (0.33)

Table B.6. (continued)

Fixed costs quintiles (2) - SGA over sales						
Q1	0.18 (0.18)	0.19 (0.16)	-0.14 (0.16)	-0.15 (0.19)	-0.27 (0.17)	-0.45* (0.26)
Q2	0.41* (0.24)	0.18 (0.19)	0.06 (0.17)	-0.21 (0.13)	-0.16 (0.15)	-0.58* (0.31)
Q3	0.66*** (0.24)	0.39** (0.17)	-0.01 (0.15)	-0.12 (0.15)	-0.03 (0.13)	-0.70** (0.31)
Q4	0.79*** (0.20)	0.29 (0.20)	0.30* (0.17)	0.24* (0.13)	0.02 (0.14)	-0.78*** (0.27)
Q5	0.60** (0.25)	-0.18 (0.23)	-0.17 (0.22)	0.09 (0.18)	-0.26* (0.15)	-0.86*** (0.29)
