Trade Liberalization, Exports and Technology Upgrading:
Evidence on the Impact of MERCOSUR on Argentinean Firms

By Paula Bustos*

This paper studies the impact of a regional free trade agreement, MERCOSUR, on technology upgrading by Argentinean firms. To guide empirical work, I introduce technology choice in a model of trade with heterogeneous firms. The joint treatment of the technology and exporting choices shows that the increase in revenues produced by trade integration can induce exporters to upgrade technology. An empirical test of the model reveals that firms in industries facing higher reductions in Brazil’s tariffs increase investment in technology faster. The effect of tariffs is highest in the upper-middle range of the firm-size distribution, as predicted by the model.

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Trade liberalization can increase productivity by inducing a better allocation of production factors or the adoption of more advanced technologies. The recent trade literature [Nina Pavcnik (2002), Marc Melitz (2003), Andrew B. Bernard et al. (2003) and James R. Tybout (2003)] has

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emphasized the first channel: trade integration reallocates market shares towards exporters, the most productive firms, increasing aggregate productivity. In this paper I show that, in addition, the resulting increase in revenues can induce exporters to invest in new technologies.

I study the impact of a regional free trade agreement on technology upgrading by Argentinean firms. To guide empirical work, I introduce technology choice in a model of trade with heterogeneous firms. In the model, more productive firms make higher revenues, therefore are the only ones that find paying the fixed costs to enter the export market profitable, like in Melitz’s (2003). In addition, only the most productive firms adopt the most advanced technology. This is because the benefit of adoption is proportional to revenues, while its cost is fixed. In this setup, a bilateral reduction in tariffs increases export revenues more than it decreases domestic revenues, inducing more firms to adopt the new technology.

I test the model in the context of a regional trade liberalization episode: MERCOSUR. I directly estimate the impact of the reduction in Brazil’s tariffs on entry in the export market and technology upgrading by Argentinean firms. Brazil’s tariffs provide a good source of arguably exogenous variation, as they fell from an average of 29% in 1991 to zero in 1995, and varied extensibly across industries. Indeed, a look at the aggregate data suggests that MERCOSUR had a strong impact on Argentina’s exports: between 1992 and 1996 exports to Brazil quadrupled while exports to the rest of the world increased only 60%.

The firm-level panel data set I analyze is uncommon in that it contains direct measures of spending in several dimensions of technology, namely computers, software, technology transfers, patents and innovation activities performed within the firm like R&D.\footnote{In addition, the survey contains a series of questions asking whether the firm performed a certain category of innovation or improvement in products or production process during the period 1992-1996 that I use to perform robustness checks.} This permits

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1 In addition, the survey contains a series of questions asking whether the firm performed a certain category of innovation or improvement in products or production process during the period 1992-1996 that I use to perform robustness checks.
to build a direct and comprehensive measure of investment in technology instead of relying on
the estimation of residuals from the production function as proxies for the level of technology.

In a first analysis of the data I check whether the sorting pattern predicted by the model
is consistent with the observed differences between exporters and non exporters operating in the
same industry. In the model, underlying productivity differences produce a sorting of firms in
three groups: the most productive firms both export and use the advanced technology, the
intermediate group exports but still uses the old technology and the least productive firms use the
old technology and serve only the domestic market. Indeed, in 1992 exporters had, on average, a
higher level of spending in technology per worker than non exporters in the same industry. The
model also predicts that during the liberalization period both old and new exporters upgrade
technology faster than non exporters, which is confirmed by the data. In particular, new
exporters were not more technology intensive than non exporters before liberalization, but
upgrade technology faster as they enter the export market during the liberalization period.

The patterns in the data described above show that there is a coincidence between entry
in the export market and technology upgrading but do not provide an answer to the question of
whether trade liberalization induced firms to adopt new technologies. Indeed, both entry in the
export market and technology upgrading could be caused by other economic reforms undertaken
in the same period if these had heterogeneous effects on firms with different characteristics.\(^2\)

Then, a second step in the empirical analysis attempts to establish causality by linking exporting
and technology adoption directly to the reduction in Brazil’s tariffs for imports from Argentina.
Note that this is a direct test of the model where both the decision to enter the export market and
to adopt a new technology are endogenous, and thus a function of tariffs.

\(^2\) For example, capital account liberalization could have made credit available for middle sized firms allowing them
to enter the export market and upgrade technology.
The model predicts that in industries where tariffs fall more, both the productivity cutoff to enter the export market and to adopt the new technology fall more. Then, to assess the impact of falling tariffs on the export decision I estimate the change in the probability that a firm enters the export market as a function of the change in Brazil’s tariffs at the industry level. I find that firms in sectors with a higher reduction in tariffs are more likely to enter the export market. The average reduction in tariffs (24 percentage points) increases the probability to enter the export market by 10 to 12 percentage points.

Next, to assess the impact of falling tariffs on the technology adoption decision I estimate the change in spending in technology as a function of the change in tariffs. I find that firms increase their spending in technology faster in industries where tariffs fall more. The average reduction in Brazil’s tariffs increases spending in technology by 0.20 to 0.28 log points. I find that the reduction in tariffs has a positive effect of similar magnitude on old and new exporters, as suggested by the within industry patterns in the data reported above.

Finally, I test the model’s prediction that that the reduction in tariffs induces firms in the middle range of the productivity distribution to enter the export market and upgrade technology, but should not affect firms in the lower and upper ranges of the distribution. I find that the reduction in Brazil’s tariffs had a stronger effect on both entry in the export market and technology upgrading in the 3rd quartile of the firm size distribution. The estimated effects on the 3rd quartile are around double the size than the average effects for all firms reported above.

The empirical identification of the effect of falling export costs on entry in the export market and technology upgrading is based on a generalized differences-in-differences estimation,

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3 As measures technology I use spending in technology, spending in technology per worker and spending in technology over sales, all produce similar results.
4 I use initial firm size measured as employment relative to the 4-digit-industry mean in 1992 as a proxy for productivity, as the survey does not provide for measures of value added nor a long enough series of investment that would permit to calculate productivity as a residual of an estimated production function.
where the sources of variation are the changes in Brazil’s tariffs for imports from Argentina across time (1996 - 1992) and across 4-digit-ISIC industries. Note that, as MERCOSUR mandates that tariffs fall to zero in all industries, I relate changes in technology spending to the initial level of Brazil’s tariffs. The focus on changes in technology differences out time-invariant industry characteristics that might be correlated with Brazil’s tariffs. The use of the initial level of Brazil’s tariffs minimizes reverse causality concerns. Still, a main potential problem is that other reforms carried out in the same period could have had heterogeneous effects on industries with different characteristics. I address this concern by showing that results are robust to controls for industry trends at the 2-digit-ISIC dissagregation level and the likely determinants of Brazilian trade policy: skill, capital intensity and the elasticity of demand of the industry at the 4-digit-ISIC dissagregation level.

The model developed in this paper builds on an extensive theoretical literature analyzing the effects of trade on technological change. In particular, it was inspired by the insight that a reduction in trade costs increases the share of firms that export and use the most advanced technology in Stephen R. Yeaple (2005). The model I present differs from Yeaple’s in that heterogeneity in exporting and technology choice is the result of ex-ante heterogeneity in productivity. To my knowledge, the model presented in this paper is the first to show that when firms are heterogeneous the presence of fixed technology adoption costs implies that the trade-

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5 For example, capital account liberalization could have benefited capital-intensive industries disproportionately. If Brazil’s trade policy was also targeting these industry characteristics, the estimates of the effects of tariffs might pick up the impact of this other policy.

6 Gene M. Grossman and Elhanan Helpman (1991) provide a comprehensive analysis of the effects of economic integration on innovation and growth; Jonathan Eaton and Samuel Kortum (2001) discuss the effect of lower barriers to trade on innovation, in particular, in their baseline model the effect of a bigger market size is counteracted by the increased competition with technologies embedded in imports, so that there is no effect of lower barriers to trade on innovation.

7 In Yeaple (2005) firms are ex-ante homogeneous, but in equilibrium all firms are indifferent between entering the export market and adopting the new technology or serving only the domestic market and using the old technology.
induced reallocations of market shares towards exporters can induce them to upgrade technology. This differential feature of the model is important to interpret the empirical findings reported above: the reduction in tariffs induced technology adoption mostly the 3rd quartile of the firm size distribution, and not only new exporters but also firms that were already exporting upgrade technology when variable trade costs fall.

The empirical work presented in this paper is related to the literature that analyzes the question of whether export market participation has a positive impact on productivity. The first studies by Sofronis K. Clerides, Saul Lach and Tybout (1998) for Colombia, Mexico and Morocco, and Bernard and Jensen (1999) for the U.S. find that exporters have higher productivity than non exporters, but this is because ex-ante more productive firms become exporters, while there are no effects of exporting on productivity. Instead, recent papers in this literature like Johannes Van Biesebroeck (2005) and Jan De Loecker (2007) find increases in productivity after firms enter the export market in Ivory Coast and Slovenia, respectively. This paper differs from this literature in that the outcome of interest is technology instead of productivity; and in that it analyzes the effect of bilateral trade liberalization on technology adoption, not the effect of exporting.

The first departure from previous literature, namely the focus on investment in technology as the outcome of interest, has the advantage of isolating a particular mechanism through which firm-productivity can improve. Earlier studies have often estimated productivity

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8 A similar approach was followed by Eric Verhoogen (2008) who develops a model where increased trade with more developed countries increases production of high quality goods and tests it in the context of Mexico’s 1994 devaluation. The mechanism generating quality upgrading in his model is the higher valuation for high quality goods of consumers in developed countries, the U.S. in this case. Instead, in this paper the analysis focuses on trade liberalization between two countries of a similar level of development, Argentina and Brazil, thus the mechanism generating technology upgrading is of a different nature: increased revenues for exporters to a country with identical homothetic preferences. Still, in the model technology upgrading can be interpreted alternatively as reducing marginal production costs or increasing quality.
as a residual in the production function. These residuals not only capture differences in technical efficiency across firms but also differences in market power, factor market distortions, or changes in the product mix, as suggested by the recent work by Lucia Foster, John Haltiwanger, and Chad Syverson (2008), Chang-Tai Hsieh and Peter Klenow (forthcoming) and Bernard, Stephen Redding and Peter Schott (forthcoming), respectively. More importantly, changes in technology not only affect productivity but can have implications for factor markets if new technologies use skilled labor more intensively. Indeed, several studies have documented increases in the relative demand for skill in developing countries during the trade liberalization period, leaving the open question of whether skill-biased technological change might have been an endogenous response to trade liberalization. This paper provides evidence for a particular channel through which increased trade can induce firms to upgrade technology, namely increased export revenues.

The second departure from existing literature, namely the estimation of the impact of a reduction in a trading partner’s tariffs on investment in technology instead of the effect of export market participation, parallels the comparative static exercise that naturally emerges from a model where both the decision to export and adopt technology are endogenous, thus each variable is a direct function of tariffs. This exercise is aimed to address the policy question of what is the effect of a reduction in a trading partner’s tariffs on technology investment, for which comparison of exporters and non exporters across time can only offer indirect evidence. Indeed, the finding that entry in the export market is not associated with increases in productivity in the absence of trade reforms can be explained by entry responding to temporary opportunities to sell in a foreign market. The opposite finding, even in the context of a trade reform, can’t be fully

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9 Pinelopi K. Goldberg and Nina Pavcnik (2007) review and discuss these studies.
attributed to it, specially in the context of simultaneous implementation of other market-oriented reforms that might have made it possible for some firms to invest in productivity improvements and thus enter the export market.

The empirical methodology implemented in this paper follows the literature measuring the effects of trade liberalization on economic outcomes through changes in tariffs. The focus of most studies has been unilateral trade liberalizations while the analysis of regional or bilateral trade liberalizations are rare. The first study of the impact of a trading partner’s reduction in tariffs using plant-level data was Daniel Trefler’s (2004) analysis of the Canada-U.S. Free Trade Agreement. To my knowledge, this paper’s analysis of MERCOSUR is the first study of the impact of a trading partner’s reduction in tariffs for a developing country. Not surprisingly, the effects of trade on technology adoption seem to be different in this context. This can be seen by comparing the results presented here with those in a contemporaneous study of the Canada-U.S. Free Trade Agreement by Alla Lileeva and Trefler (forthcoming). Their finding that the reduction in U.S. tariffs only induced productivity increases in the least productive new entrants in the export market for the case of Canada contrasts with the findings for Argentina where the reduction in Brazil’s tariffs induced technology upgrading mostly in the 3rd quartile of the firm-size distribution and not only in new but also in old exporters. As I discuss in the theoretical section of the paper, the result that old exporters upgrade technology when trade costs fall only obtains when the costs of technology adoption are high (relative to fixed exporting costs) which is more likely to be the case in developing countries.

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10 This literature includes studies of the impact of trade liberalization on inequality like Orazio Attanasio, Goldberg and Pavcnik (2004) for Colombia, Petia Topalova (2005) for India, and the study of the impact of trade liberalization on productivity in Colombia by Ana Fernandes (2007).
The remaining of the paper is organized as follows. The next section presents the theoretical model and derives the empirical predictions on the effects of trade liberalization on entry in the export market and technology upgrading. Section II describes the trade liberalization episode and the data set. Section III presents the empirical strategy and tests the predictions of the model. Section IV concludes.

I. Theory
This section develops a simple model of the decision to enter the export market and upgrade technology by heterogeneous firms. I consider the case of two symmetric countries engaging in bilateral trade liberalization. Each economy consists of a single monopolistically competitive industry where firms produce differentiated products under increasing returns to scale, and using a single factor of production, labor, as in Paul Krugman (1979). Firms are heterogeneous in productivity, face fixed exporting costs as in Melitz (2003), and can choose to increase their productivity by paying a fixed technology adoption cost, as in Yeaple (2005).

A. Set up of the Model
Each country is endowed with \( L \) units of labor used to produce differentiated products in a single industry. The symmetry assumption ensures that wages, which are the numeraire, and all aggregate variables are the same for both countries. I present the discussion from the point of view of the home country.

Entry
The supply side is characterized by monopolistic competition. Each variety is produced by a single firm, and there is free entry into the industry. Firms are heterogeneous in their productivity
in the sense that marginal labor costs vary across firms using the same technology.\textsuperscript{11} This idiosyncratic component of labor productivity is indexed by $\varphi$, that also indexes firms and varieties. To enter the industry in a given country, firms pay a fixed entry cost consisting of $f_e$ units of labor. Entrants then draw their productivity from a known Pareto cumulative distribution function $G(\varphi) = 1 - \varphi^{-k}$ with $k > 1$.

\textit{Technology}

After observing their productivity firms decide whether to exit the market or stay and produce. Firms produce varieties using a technology that features a constant marginal cost $(1/\varphi)$ and a fixed cost $(f)$, both in terms of labor. Firms can choose to upgrade their technology in the following sense: by paying an additional fixed cost they can reduce their marginal cost of production. This can be represented as a choice between two different technologies $l$ and $h$, where $h$ features a higher fixed cost $(\eta f)$ and a lower marginal cost $[1/(\gamma \varphi)]$. The resulting total cost functions under each technology are:

$$
TC_l(q, \varphi) = \left( f + \frac{q}{\varphi} \right)
$$

$$
TC_h(q, \varphi) = \left( f\eta + \frac{q}{\gamma \varphi} \right)
$$

where $\eta > 1$ and $\gamma > 1$. Then, in this setup, there is a part of firm productivity that is the result of luck but firms can also take actions to increase their productivity. A simple interpretation would be that before entering an industry firms engage in product development, but the value of that product/its marginal production cost is revealed only after it has been developed and thus the cost of product development is sunk. At the production stage, firms can take actions to increase

\textsuperscript{11} Alternatively, heterogeneity in productivity can be interpreted as quality: more productive firms produce a good of higher quality, in the sense that consumers are willing to pay more for the same amount of the good.
the quality of the product or further reduce its marginal cost, by paying a higher fixed production
cost every period. Finally, in every period there is an exogenous probability of exit (δ).

Serving the Foreign Market

After entry, a firm can choose to export, in which case it must incur an additional fixed cost \( f_x \). In
addition, exported goods are subject to per-unit iceberg trade costs, so that \( \tau \) units need to be
shipped for 1 unit to make it to the foreign country.

Demand

Preferences across varieties have the standard CES form, with an elasticity of substitution
\( \sigma = 1/(1 - \rho) > 1 \). These preferences generate a demand function
\[ q(\omega) = EP^{\sigma - 1} [p(\omega)]^{-\sigma} \]
for every variety \( \omega \), where \( p(\omega) \) is the price of each variety. \( P = \left[ \int_0^M p(\omega)^{1 - \sigma} d\omega \right]^{1/\sigma} \)
is the price index of the industry, \( M \) is the number (measure) of existing varieties and \( E \) is the aggregate level
of spending in the country.

B. Firm Behavior

Profit Maximization

Under CES preferences the profit maximizing price is a constant markup over marginal costs.
Then, a firm with productivity \( \phi \) using technology \( l \) charges the price
\[ p^d_l(\phi) = 1/(\rho \phi) \]
in the
domestic market and a higher price in the export market \( p^e_l(\phi) = \tau/(\rho \phi) \). If instead the firm
uses technology \( h \), it charges lower prices in both markets: \( p^d_h(\phi) = 1/(\rho \phi) \) and
\[ p^e_h(\phi) = \tau/(\rho \phi) \].

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To make the joint decision of whether to enter the export market and whether to adopt technology \( h \), firms compare the total profit of each of the four possible choices, which are described below.

Profits if only serving the domestic market and using technology \( l \):
\[
\pi_d^l(\phi) = \frac{1}{\sigma} E(P\rho)^{\sigma-1} \phi^{\sigma-1} - f
\]

Profits if only serving the domestic market and using technology \( h \):
\[
\pi_d^h(\phi) = \frac{1}{\sigma} E(P\rho)^{\sigma-1} \phi^{\sigma-1} \gamma^{\sigma-1} - f\eta
\]

Profits if also exporting and using technology \( l \):
\[
\pi_x^l(\phi) = (1 + \tau^{1-\sigma}) \frac{1}{\sigma} E(P\rho)^{\sigma-1} \phi^{\sigma-1} - f - f_x
\]

Profits if also exporting and using technology \( h \):
\[
\pi_x^h(\phi) = (1 + \tau^{1-\sigma}) \frac{1}{\sigma} E(P\rho)^{\sigma-1} \phi^{\sigma-1} \gamma^{\sigma-1} - f\eta - f_x
\]

Note that the assumption that both countries are identical and trade costs are symmetric implies that the price index \((P)\) and the expenditure level \((E)\) in foreign are the same as at home. Exporting and technology choices are represented in Figure 1, where the four possible profits are depicted as a function of firm's productivity.\(^{12}\) The equilibrium depicted is obtained when \( \phi^x < \phi^h \), where \( \phi^x \) is defined as the level of productivity above which a firm using technology \( l \) finds exporting profitable \( [\pi_x^l(\phi^x) = \pi_x^l(\phi^x)] \) and \( \phi^h \) is defined as the level of productivity above which an exporter finds adoption of technology \( h \) profitable \( [\pi_x^h(\phi^h) = \pi_x^h(\phi^h)] \). In Appendix A I show that in this equilibrium firms sort into four different

\(^{12}\) More precisely a transformation of firm's productivity: \( \phi^{\sigma-1} \).
groups: the least productive firms \((\varphi < \varphi^*)\) exit, the low productivity firms \((\varphi^* < \varphi < \varphi^h)\) only serve the domestic market and use technology \(l\), the medium productivity firms \((\varphi^i < \varphi < \varphi^h)\) still use technology \(l\) but also export, and the most productive firms \((\varphi^h < \varphi)\) both export and use technology \(h\).

Note that in Figure 1 using technology \(h\) and only serving the domestic market is always dominated by some other choice. Note also that there is a range of productivity levels where exporting is profitable but adopting technology \(h\) is not, so that the marginal exporter uses technology \(l\). I focus in this case \((\varphi^i < \varphi^h)\) in what follows and provide the necessary parameter restrictions for this ordering of cutoffs to apply. The opposite case \((\varphi^i > \varphi^h)\) is one where the equilibrium features no exporters using the low technology, which is inconsistent with the empirical findings I report in the next section.

**Figure 1**

**Exporting and Technology Choices**
To solve for the industry equilibrium it is useful to state the conditions for exit, entry in the export market and technology adoption as a function of the exit cutoff, which I do next.

Exit

For the least productive firms profits are highest when using technology $l$ and only serving the domestic market. Then the exit cutoff $\phi^*$ is defined by:

$$
\pi^d_i(\phi^*) = 0 \iff \frac{1}{\sigma} E(P\rho)^{\sigma-1}(\phi^*)^{\sigma-1} - f = 0
$$

Exporting

The marginal exporter uses technology $l$. Then $\phi^*$ can be expressed as a function of $\phi^*$ using $\pi^d_i(\phi^*) = \pi^e_i(\phi^*)$ and the zero profit condition for the marginal firm (eq. 1):

$$
\phi^* = \phi^* \tau \left( \frac{f}{f_1} \right)^{\frac{1}{\sigma-1}}
$$

note that as long as $\tau(f_1 / f)^{\frac{1}{\sigma-1}} > 1$, $\phi^* > \phi^*$. Thus, only the most productive firms export.
Technology Choice

The marginal firm adopting technology $h$ is an exporter. Then the adoption cutoff ($\phi^h$) is defined by:

$$\pi^*(\phi^h) - \pi^*_i(\phi^h) = 0 \iff (\phi^{\sigma-1} - 1)(1 + \tau^{1-\sigma})\frac{1}{\sigma} E(P)\gamma^{\sigma-1}(\phi^h)^{\gamma-1} = f(\eta - 1)$$

The benefit of using technology $h$ (the L.H.S. of the equation above) is that the firm makes higher revenues, as demand is elastic ($\sigma > 1$). The cost of using technology $h$ (the R.H.S. of the equation above) is its higher fixed cost. Note that this cost is the same for all firms while the benefit is increasing in productivity. This is why technology choice is characterized by a cutoff productivity level $\phi^h$ above which all firms use technology $h$. Next, $\phi^h$ can be expressed as a function of $\phi^*$ by substituting the zero profit condition for the marginal firm (eq. 1) in the equation above:

$$\phi^h = \phi^* \left( \frac{1}{(1 + \tau^{1-\sigma})^{\frac{1}{\gamma-1}}} \right)\left( \frac{\eta - 1}{\phi^{\sigma-1} - 1} \right)^{\frac{1}{\gamma-1}}$$

Note that the share of active firms adopting technology $h$ [($\phi^h / \phi^*$)] is higher the lower are variable trade costs. This is because a reduction in trade costs increases the total revenues of exporters relative to those of the marginal firm which only serves the domestic market.\(^\text{13}\) By comparing equations (2) and (3) we can see that the parameter restriction required for $\phi^h > \phi^*$ is that technology adoption costs are high enough relative to fixed exporting costs:

$$\frac{\phi^h}{\phi^*} = \left( \frac{\tau^{1-\sigma} \gamma^{\sigma-1}}{(1 + \tau^{1-\sigma})^{\frac{1}{\gamma-1}}} \frac{\eta - 1}{\gamma^{\sigma-1} - 1} \right)^{\frac{1}{\gamma-1}} > 1$$

\(^\text{13}\) Indeed, in Appendix C I show that this result requires that the marginal firm is a non-exporter, that is $\tau(f_1/f)^{\frac{1}{\gamma-1}} > 1$. This is implicitly assumed in the zero profit condition for the marginal firm (eq. 1) used to derive equation (3).
C. Industry Equilibrium

The equilibrium price \((P)\), number of firms \((M)\) and the distribution of active firms' productivities in the economy are determined by the free entry condition. Free entry requires that the sunk entry cost equals the present value of expected profits:

\[
f_e = [1 - G(\phi^*)] \frac{1}{\delta} \bar{\pi}
\]

where \(1 - G(\phi^*)\) is the probability of survival and \(\bar{\pi}\) are per-period expected profits of surviving firms. \(\bar{\pi} = \bar{\pi}_d + p_x \bar{\pi}_s\) where \(\bar{\pi}_d\) are expected profits from domestic sales, \(p_x = [1 - G(\phi^*)]/[1 - G(\phi^*)]\) is the probability of exporting conditional on surviving and \(\bar{\pi}_s\) are expected exporting profits. Then, to solve for the free entry condition (eq. 4) we need to solve for expected profits \(\bar{\pi}\). The derivations are detailed in Appendix A:

\[
\bar{\pi} = \left( \frac{\sigma^{-1}}{k - (\sigma - 1)} \right) f \Delta
\]

\[
\Delta = \left\{ 1 + \left( \frac{f_x}{f} \right)^{\frac{1}{\sigma}} \right\} \frac{f_x}{f} + \left[ \frac{\eta - 1}{(1 + \tau^{1-\sigma})(\gamma^{\sigma-1} - 1)} \right]^{\frac{1}{\sigma}} (\eta - 1)
\]

By substituting the solution for expected profits (eq. 5)\(^\text{14}\) in the free entry condition (eq. 4) we can solve for the exit cutoff:

\[
\phi^* = \left[ \frac{f}{\delta} \left( \frac{\sigma^{-1}}{k - (\sigma - 1)} \right) \Delta \right]^{\frac{1}{\eta}}
\]

\(^\text{14}\) Note that for expected profits to be positive we need to impose the parameter restriction: \(k > \sigma - 1\).
By substituting the solution for the exit cutoff (eq. 6) in eqs. 2 and 3 a solution for the exporting and technology adoption cutoffs can be obtained:

\[
\varphi^* = \left[ \frac{f}{\delta f_e} \left( \frac{\sigma - 1}{k - (\sigma - 1)} \right) \right]^{\frac{1}{\Delta^l}} \left( \frac{f_y}{f} \right)^{\frac{1}{\Delta^l}}
\]

\[
\varphi^b = \left[ \frac{f}{\delta f_e} \left( \frac{\sigma - 1}{k - (\sigma - 1)} \right) \right]^{\frac{1}{\Delta^l}} \left( \frac{1}{(1 + \tau^{1-\sigma})^{\frac{1}{\Delta^l}}} \right) \left( \frac{\eta - 1}{\epsilon^{\sigma-1} - 1} \right)^{\frac{1}{\Delta^l}}
\]

Finally, welfare is determined by the inverse of the price index, which can be obtained by substituting the exit cutoff (eq. 6) in the zero-profit condition for the marginal firm (eq. 1):

\[
P = \frac{1}{\rho} \left( \frac{\sigma f}{L} \right)^{\frac{1}{\Delta^l}} \left[ \frac{f}{\delta f_e} \left( \frac{\sigma - 1}{k - (\sigma - 1)} \right) \right]^{\frac{1}{\Delta^l}} \Delta^l
\]

**Discussion**

To interpret the solution for expected profits in eq. 5 note that \( f \Delta \) can be written as:

\[
f \Delta = f + p_s f_s + p_h (f_h - f)
\]

where \( p_s = (\varphi^s / \varphi^*)^{-k} \) and \( p_h = (\varphi^b / \varphi^*)^{-k} \) are the fraction of surviving firms that export and adopt the high technology, respectively. Then, expected profits are proportional to expected fixed costs \( (f \Delta) \). It is straightforward to show that in the simple case of a closed economy with only one technology the solution for expected profits is the same as in eq. 5 but with \( \Delta = 1 \). Then, expected profits are proportional to the variable profits of the marginal surviving firm, which must be equal to \( f \). In the open economy, with probability \( p_s \) the firm becomes an exporter, and in that case expected profits are augmented in proportion to \( f_s \), the variable exporting profits of the marginal exporter. Finally, with probability \( p_h \) the firm adopts the high technology, in which case expected profits are augmented in proportion to the variable adoption profits of the
marginal adopters which are \((f_h - f)\). Note that as a reduction in variable trade costs increases the fraction of firms that export \(p_x\) and the fraction of firms adopting the high technology \(p_h\), expected profits increase.

**D. Bilateral Trade Liberalization**

In this section I analyze the impact of bilateral trade liberalization on entry in the export market and technology upgrading. I show that a reduction in trade costs increases export revenues, inducing more firms to enter the export market and upgrade technology. This increases expected profits, inducing more entry into the industry. Increased entry reduces the price index and thus firms only serving the domestic market loose revenues. As a result, the least productive firms make negative profits and exit.

More formally, I show in Appendix B that when variable trade costs \((\tau)\) fall, and not all firms export \((\tau^{\sigma-1} f_x > f)\):

1. **The fraction of surviving firms that export**, \(p_x = \left(\varphi^* / \varphi^*\right)^k\), and the fraction of surviving firms that use technology \(h\), \(p_h = \left(\varphi_h^* / \varphi^*\right)^k\), increase.\(^{15}\)

2. **Expected profits increase**, that is \(\partial \widehat{\pi} / \partial \tau < 0\).

3. **The price index falls**, that is \(\partial P / \partial \tau > 0\).

4. **The exit productivity cutoff increases**, that is \(\partial \varphi^* / \partial \tau < 0\).

5. **The productivity cutoff for exporting decreases**, that is \(\partial \varphi_x / \partial \tau > 0\).

6. **The productivity cutoff for adopting technology \(h\) decreases**, that is \(\partial \varphi_h / \partial \tau > 0\).

**Discussion**

\(^{15}\) This can be directly seen in eqs. 2 and 3.
The new result in the model is that the reduction in variable trade costs induces more firms to upgrade technology (Result 6). What makes adoption of the new technology profitable for the most productive exporters is the increase in total revenues. Still, it is important to note that this is not a market size effect: an increase in market size as represented by an increase in $L$ does not affect revenues nor the technology adoption cutoff. Instead, the result is due to the asymmetric effect of trade liberalization in models of heterogeneous firms with fixed exporting costs: while firms serving only the domestic market loose revenues, exporters see their revenues increase.

Indeed, this result requires that domestic revenues fall less than export revenues increase. I show in Appendix C that this can never be the case when the marginal firm is an exporter. In that case, as $\tau$ falls free entry induces the price index to fall enough to make the profits of the marginal firm equal to zero. If this firm is an exporter, the price index must fall enough to make the reduction in domestic profits completely offset the increase in export profits.

An alternative intuition for this result is that as countries engage in bilateral trade liberalization, firms lose domestic revenues because there are more foreign firms and increased foreign sales, but gain export revenues. The second effect dominates as long as exporters can serve the foreign market but face the entry of only a fraction of foreign firms.

II. Context and Data

A. Trade Liberalization

In this section I describe the regional and unilateral trade liberalization policies undertaken in Argentina at the beginning of the 1990’s. Although these policies had started to be discussed in the late 1980’s, the depth and pace of the reforms implemented in 1991 were largely unexpected. The newly elected president had promised populist policies during the campaign, namely a

\[16\] The benefit of technology adoption is proportional to revenues while its cost is fixed.
widespread increase in wages, but his government implemented a set of market oriented reforms. Many observers believed that the newly built consensus for the reforms was mostly due to the 1989 and 1990 hyperinflations, and the crisis in the socialist bloc. In particular, political arguments favoring MERCOSUR in Argentina and Brazil were based in the view that after the fall in the Berlin Wall the world would be organized in regional blocks, as the recent emergence of NAFTA and creation of the EU suggested.\(^{17}\)

Argentina started reducing import tariffs with respect to the rest of the world before MERCOSUR was launched, in the context of debt-related negotiations with the World Bank and the IMF. Between October 1988 and October 1991 there were 11 major revisions of trade policy, often related to changes in macroeconomic policy aimed at controlling hyperinflation. By October 1991, the average nominal tariff was 12%, ranging from 0% to 35%, where rates were increasing in the value-added of production of each good. Manufactures were concentrated in the range of 5% to 22%. Almost all import licenses were eliminated, with the exception of the automobile industry. Finally, in October 1993 imports of new capital goods were exempted of tariffs.

MERCOSUR was established by Argentina, Brazil, Paraguay and Uruguay in March 1991. The agreement established generalized, linear and automatic reductions in tariffs, and the adoption of a common tariff with third countries. The tariff reductions were generalized in the sense that the same reduction relative to the most-favored nation (m.f.n.) tariff rates was to be applied to all goods. They were to be implemented gradually according to a semi-annual timetable starting by a 54% reduction in December 1991, and finishing at 100% in December

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\(^{17}\) For a discussion of the policy debates in Argentina and Brazil during the period of launching of MERCOSUR see Jorge Campbell, Ricardo Rozemberg and Gustavo Svarzman (1999).
1994. This new agreement was in sharp contrast with the regional integration treaty signed in 1988, where reductions in tariffs were gradually negotiated sector by sector and free trade was to be achieved in 10 years.

The Customs Union was established in 1995 with the adoption of a Common External Tariff (CET), with an average level of 12%. Tariffs varied between 0 and 20% across industries. Inputs and materials had the lowest tariffs, followed by semi-finished industrial goods, capital and IT goods, and final goods. There were exceptions to internal free trade for a limited number of products, special regimes for sugar and automobiles and some products faced tariff rates different from the CET.[Insert Table 1]

MERCOSUR seems to have had a big impact on Argentinean exports. Between 1992 and 1996, exports to Brazil quadrupled, while exports to the rest of the world only increased 60%. As a result, growth in exports to Brazil explains 50% of the growth in total exports during this period. This might be related to deep reduction in Brazil’s tariffs during this period. Table 1 reports summary statistics for m.f.n tariffs at the 4-digit-ISIC industry level of aggregation in the period under study. The first row reports the level of Brazil’s m.f.n. tariffs in 1991 which are the baseline for the MERCOSUR tariff reductions that started in December 1991. The average

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18 The timetable of reductions relative to m.f.n. rates was: 54% by December 1991, 61% by June 1992, 68% by December 1992, 75% by December 1993, 82% by December 1993, 89% by June 2004 and finally 100% by December 1994.

19 According to Julio Berlinski et al. (2006) the common external tariffs for capital goods (14%) and information technology and telecommunication (16%) were the most difficult to agree upon. Argentina favored low tariffs while Brazil wanted higher protection. Thus, national tariffs were to converge to the CET by 2001 for capital goods and 2006 for IT goods, from above in the case of Brazil and from below in the case of Argentina.

20 The source of the tariff data is UNCTAD-TRAiNS. Tariffs for each 4-digit-ISIC-industry are obtained as a weighted average of the 9-digit-HS-products within each 4-digit-ISIC-industry, where the weights are given by imports of each product. Thus, when computing Brazil’s m.f.n tariffs in 1992 weights for each product within a 4-digit industry are based on Argentina’s exports to Brazil of that product in each year. An alternative is to obtain 4-digit-ISIC-industry as simple averages of m.f.n tariffs for 9-digit-HS-products within each industry, but these give similar results as their correlation is 0.975.
reduction in Brazil’s tariffs faced by Argentinean firms between December 1991 and December 1994 was 29 p.p. Tariff reductions varied extensively across industries, as initial m.f.n tariffs varied between 84 p.p. and 0 p.p.. As the panel of firms I analyze covers the period 1992-1996, I use the level of Brazil’s m.f.n tariffs in 1992 as the baseline for the calculation of tariff reductions in the period 1992-1996. These are on average 24 p.p., slightly lower than 1991 tariffs but reflect a similar variation across industries, as their correlation is 0.97.

As m.f.n. tariffs in Argentina were already low before MERCOSUR was launched, the baseline for the reduction in Argentina’s tariffs for imports from Brazil was only 13 percentage points on average (Table 1). Still, there was significant variation in tariffs across 4-digit-ISIC industries, from 0 to 22 p.p. Surprisingly, imports from Brazil grew exactly at the same rate as imports from the rest of the world during this period (60%).

As Argentina’s unilateral trade liberalization occurred before the period under study, between 1992 and 1996 Argentina’s average import tariffs with respect to the rest of the world increased slightly (1 p.p.). Still, there were changes in tariffs in both directions, from -10 p.p. to 14 pp. across 4-digit-ISIC industries. The modifications on import tariffs during this period are partly related to the convergence to the CET, that partly reflected the structure of protection in Brazil.\(^{21}\)

In addition, Table 1 reports average m.f.n. input tariffs for Argentina as these are used for robustness checks in the empirical analysis of the impact of Brazil’s tariffs on entry in the export market and technology upgrading. The input tariff for each industry is computed as a weighted average of the tariffs of all inputs used, where the weights are based on the cost share of each input obtained from the input-output matrix of Argentina, as described in Appendix D.

\(^{21}\) Berlinsky et al. (2006) and Won Chang and L. Alan Winters (2002) provide a more detailed discussion of Argentina and Brazil’s trade policy measures in the 1990’s.
The baseline m.f.n rates for Argentina’s input tariff reductions w.r.t Brazil were smaller than the output tariffs reported above, with an average level of 11 p.p. in 1992. Similarly, the changes in Argentina’s input tariffs w.r.t. the world were smaller than the output tariffs, ranging from -3 to 6 p.p.

Finally, an important point to note is that the start of MERCOSUR tariff reductions respect to m.f.n. rates, December 1991, just precedes the period under study 1992-1996. Still, exports seem to have reacted to tariff declines with a lag. The data on aggregate Argentinean industrial exports to Brazil shows that these started growing in 1993. Thus, it is likely that the relevant overall tariff reductions in the period 1992-1996 are the full 100% reduction over m.f.n. rates between December 1991 and 1994 and not the 32% remaining reduction that occurred between December 1992 and 1994. Thus, in the empirical analysis I set the change in Brazil’s tariffs w.r.t Argentina between 1992 and 1996 to minus the level of Brazil’s m.f.n tariffs in 1992. Similarly, I set the change in Argentina’s tariffs w.r.t Brazil between 1992 and 1996 to minus the level of Argentina’s m.f.n tariffs in 1992. Note that the application of a 100% or 32% tariff reduction w.r.t. m.f.n tariffs in 1992 does not affect the estimation of the average impact of tariffs on entry in the export market or technology upgrading as in the first case the estimated coefficient is 0.32 times smaller but the average change in tariffs is (1/0.32) times bigger. It does affect the interpretation of the results, though, as the implied responses of entry in the export market and spending in technology to a given tariff change are 0.32 times smaller when considering the full 100% reduction. Then, the reported estimates can be considered as a lower bound.

Brazil’s Trade Policy

[22 For example, if the change in Brazil’s tariffs is set to minus the level of m.f.n tariffs in 1992 multiplied by 0.32, estimated coefficients are 1/0.32 times bigger but then the average reduction in Brazil’s tariffs in the period is 0.32 times smaller, thus the estimated effect of the average reduction of tariffs is the same.]
As the source of identification of the effect of tariff reductions on entry in the export market and technology upgrading are the differences across industries in the level of m.f.n tariffs in Brazil in 1992, it is important to discuss Brazil’s trade policy in more detail.

Like Argentina, Brazil implemented a program of unilateral trade liberalization between 1988 and 1994. Berlinski et al. (2006) note that the tariff structure in 1988 was based on the tariffs implemented in 1957 under the import substitution policy. They argue that the first reforms implemented in 1988-89 did not have significant effects on the degree of protection of the domestic industry as NTBs, which were the main instrument of protection, were not modified. Instead, after a new government took power in march 1990 NTBs were eliminated and tariffs were reduced gradually according to a timetable ending in January 1994. The new tariffs would vary between 0 and 20%, except for a few goods facing 30-35% tariff rates.23

Brazil’s m.f.n tariff rates in 1992 reflect a transition between the old and new tariff structure. As a result, they display tariff rates above 30 p.p. for some unskilled, labor-intensive industries protected under the import substitution policy like toys, textiles and rubber and also for skill-intensive industries that were protected under the new policy like domestic appliances, office accounting and computing and the car industry. Possibly as a result, the correlation between Brazil’s tariffs in 1992 and an exogenous measure of skill intensity of the industry24 is very low (-0.002). Instead, tariffs are negatively correlated with a measure of capital intensity (-0.21), suggesting that Brazil protected labor-intensive industries. As the omission of industry characteristics that are correlated with Brazil’s trade policy might induce biases in the estimation

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23 According to Berlinski et al. (2006) the 0% tariffs corresponded to commodities and “exportables”, 10% for agricultural products and their derivatives, 10,15 and 20% for products using basic inputs with 0% tariffs and 20% for the rest of the products. The main exceptions to the general rule were IT goods with a 35% tariff, domestic appliances (30%) and the car industry (35% tariff).
24 I use measures of average capital and skill intensity in the industry in the U.S. in the 1980’s obtained from the NBER productivity database (see Appendix D for details).
of the impact of the reduction in Brazil’s tariffs on entry in the export market and technology upgrading, I include in the regressions 2-digit-ISIC-industry dummies that absorb part of the correlation between changes in tariffs and industry characteristics. After including 2-digit-ISIC-industry dummies the correlation between capital intensity and tariffs falls to -0.06, although the correlation between tariffs and skill intensity increases to 0.06. Thus, in addition to including 2-digit-ISIC-industry dummies I control for measures of capital, skill intensity and the elasticity of demand.\footnote{I use the elasticity of substitution in the industry as estimated by Christian Broda and David Weinstein (2006). The correlation of the elasticity of demand with tariffs is low: 0.05 and 0.06 with controls for 2-digit-ISIC-industry dummies.}

**B. Firm-Level Data**

The data I analyze comes from the *Encuesta Nacional de Innovación y Conducta Tecnológica de las Empresas Argentinas (ENIT)* [National Survey on Innovation and Technological Behavior of Industrial Argentinean Firms] conducted by the *Instituto Nacional de Estadística y Censos (INDEC)*, the Argentinean government statistical agency. The survey covers the period 1992-1996 and was conducted in 1997 over a sample of 1,639 industrial firms.

The sample is representative of firms owning establishments with more than 10 employees, and is based on 1993 census data. Although according to the census only 15% of establishments had more than 10 employees, they represented 90,7% of the value of output, 90,9% of industrial value added, 87,9% of employment and 94,1% of the wage bill.\footnote{The sample is the same as the one used for the *Encuesta Industrial Annual*, the standard yearly industry survey used to compute Industrial GDP. A description of the sampling methodology of *Encuesta Industrial Annual* is available at INDEC’s website: www.indec.mecon.ar.}

As the survey was conducted in 1997, it does not contain information on firms that were active in 1992 and exited afterwards. I focus my analysis on a balanced panel of 1,380 firms.
present both in 1992 and 1996 for which there is information on sales, employment and belong to 4-digit-ISIC industries with information on Brazil’s tariffs.

The survey contains information on several dimensions of spending on technology upgrading. Firms upgrade technology by performing various innovation activities like internal R&D, paying for technology transfers and buying capital goods that embody new technologies; and with different purposes like changing production processes, products, organizational forms or commercialization. I constructed a measure of spending on technology (ST) that includes the following: spending on computers and software; payments for technology transfers and patents; and spending on equipment, materials and labor related to innovation activities performed within the firm.27

The survey contains information on ST for all years in the period 1992-1996, while information on all the rest of the variables (sales, exports, imports, employment by education, investment) is only available for the years 1992 and 1996.

The survey also contains some binary measures of technology adoption: a list of 9 yes/no questions asking whether the firm performed a certain category of innovation or improvement in products or production process during the period 1992-1996. As an example, one of these categories is: “product differentiation” and another “machinery and equipment associated to new production process”. I use this information to construct an innovation index equal to the fraction of categories for which the firm gave positive answers. A detailed description of the questions is contained in Appendix D.

The main measure of technology I use in the empirical analysis is technology spending while the binary measures of technology are used to perform robustness checks. I think

27 Like R&D, adaptation of new products or production processes, technical assistance for production, engineering and industrial design, organization and commercialization.
technology spending is a better measure of technology for two reasons. First, the information has a panel structure that can be used to control for unobserved firm and industry characteristics. Second, it is a more objective measure in the sense that it does not depend on the interpretation of what an improvement or innovation is.

Finally, another unusual feature of the survey is that it contains information on employment by education. I use this information to construct measures of employment in primary school equivalents, skill intensity and sales per worker as described in Appendix D.

Table D.1 in Appendix D contains summary statistics by export status for the main variables of interest for the initial year in the data, 1992.

C. Industry-Level Data

In the empirical section I use controls for 4-digit-ISIC industry characteristics that might be correlated with changes in tariffs. First, average capital and skill intensity in the industry in the U.S. in the 1980’s obtained from the National Bureau of Economic Research (NBER) productivity database (see Appendix D for details). I also use the elasticity of substitution in the industry as estimated by Broda and Weinstein (2006). Finally, data on exports from Argentina to Brazil in the years 1992 and 1996 were obtained from the U.N. COMTRADE database. This information is aggregated at the 4-digit-ISIC industry.

III. Empirics

In this section I test the predictions of the theoretical model developed in section I. First, I check whether the sorting pattern of firms into exporting and technology use predicted by the model is consistent with the observed characteristics of exporters and non-exporters in the same 4-digit-
ISIC industry. Second, I test the main predictions of the model: that a reduction in variable trade costs causes entry in the export market and technology upgrading. To establish causality, I use the differential changes in Brazilian tariffs across 4-digit-ISIC industries to show that firms are more likely to enter the export market and upgrade technology in industries where tariffs fell more.

A. Within-Industry Patterns in the Data

In the model, underlying productivity differences produce a sorting of firms into three groups: the low productivity firms only serve the domestic market and use the low technology, the medium productivity firms still use the low technology but also export, and the most productive firms both export and use the high technology. In this setting a reduction in variable trade costs increases exporting revenues inducing firms in the middle-range of the productivity distribution to enter the export market and upgrade technology.

Figure 2 illustrates the effects of trade liberalization for firms in each part of the productivity distribution. The upper line represents productivity cutoffs to adopt the high technology and to enter the export market before liberalization \((\phi^*_0, \phi^h_0)\), while the lower line represents the cutoffs after liberalization \((\phi^*_1, \phi^h_1)\). Within the group of firms that were already exporting before liberalization \((\phi^*_0 < \phi)\) those in the upper range of productivity \((\phi^*_0 < \phi)\) were already using technology \(h\), while firms in the range \(\phi^*_0 < \phi < \phi^h_0\) adopt it only afterwards. Within the group of firms that enter the export market after liberalization \((\phi^*_1 < \phi < \phi^*_0)\), those in the upper range \((\phi^h_1 < \phi < \phi^*_1)\) enter the export market and adopt the new technology, while those in the lower range \((\phi^*_1 < \phi < \phi^h_1)\) enter the export market but keep the old technology.
To check whether the sorting pattern depicted in Figure 2 and the parameter restrictions required to obtain it are consistent with the data I divide firms into three groups: continuing exporters,\(^{28}\) new exporters,\(^{29}\) and never exporters\(^{30}\) and compute differences in characteristics for firms operating within the same 4-digit-ISIC industry.

[Insert Table 2]

Table 2 reports that, on average, continuing exporters have a 0.33 log points higher level of spending in technology per worker a than never exporters in 1992. This is consistent with at least a fraction of them already using the high technology before liberalization. Interestingly, they increase spending in technology 0.28 log points faster than never exporters during the

\(^{28}\) Firms that were already exporting in 1992.
\(^{29}\) Firms that export in 1996 but were not exporting in 1992.
\(^{30}\) Firms that do not export in 1992 nor 1996.
liberalization period (1992-1996), which is consistent with a fraction of them adopting the high technology after liberalization.

Firms that would enter the export market after liberalization were not significantly more technology intensive than never exporters in 1992 (Table 2). In contrast, after liberalization these new exporters become more technology-intensive than firms that do not export, increasing their spending in technology per worker 0.37 log points faster between 1992 and 1996.

The patterns in the data described above show that there is a coincidence between entry in the export market and technology upgrading, but can’t establish whether it is expanded export opportunities that cause technology adoption, vice versa, or whether both are caused by a third factor. Some alternative explanations for the results in Table 2 can be ruled out: as these are based on comparisons of exporters and non exporters within industries, they are robust to macroeconomic shocks that affect all firms equally (an example could be exchange rate appreciation) or to shocks that affect all firms within an industry (an example could be fast technological change in a particular industry). Still, the fact that within each sector exporters and new exporters are upgrading technology faster than other firms could reflect other shocks that affect middle and high productivity firms differentially. This is particularly plausible in a context where several reforms were implemented at the same time. For example, capital account liberalization, that could facilitate access to credit to finance technology upgrading and entry in foreign markets to medium and big firms but not to small firms in the presence of credit constraints. Then, the next step in the empirical analysis attempts to establish causality by linking exporting and technology adoption directly to the reduction in Brazil’s tariffs for imports from Argentina.
B. The Impact of the Reduction in Brazil’s Tariffs: Identification Strategy

The empirical identification of the effect of the fall in variable export costs on entry in the export market and technology upgrading by Argentinean firms is based on the differential reductions in Brazilian tariffs for imports from Argentina across 4-digit-ISIC industries.

This source of identification has two features that make it likely to be exogenous with respect to the outcomes analyzed, changes in export status and changes in spending in technology between 1992 and 1996. First, the tariff reductions were programmed in 1991, and reach a level of zero for all industries in 1995. Thus changes in tariffs are predetermined by the 1991 m.f.n tariff levels in Brazil. Second, the 1991 m.f.n import tariffs of Brazil are the same for Argentina and the rest of the world and are therefore unlikely to be targeted to industry characteristics particular to Argentina, whose share of Brazil’s trade was only 7.7%. As changes in tariffs are predetermined, they are unlikely to be driven by political pressures arising from the effects of liberalization in Brazil or Argentina, or by contemporaneous shocks to industrial performance. As they respond to Brazil’s worldwide trade policy, it is also unlikely that results are driven by Brazilian tariffs being initially high in industries where Argentina has a comparative advantage.

Although the points above address the reverse causality problem, Brazil’s initial tariff structure is certainly not random. As discussed above, Brazil’s trade policy is correlated with some industry characteristics, and omitting them could be an important source of bias. Thus, I estimate all the equations in first differences, so that constant industry characteristics are differenced-out. Still, if industries with different initial characteristics are on different trends,  

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31 Except for the automobile and sugar industries. In the results presented in this section, 1996 tariffs are still set to zero for these two industries, to avoid endogeneity problems in using the actual 1996 tariffs. As a robustness check, all the results presented in this section have been replicated for the sample of firms excluding these sectors.

32 Argentina’s share on Brazil’s imports rose to 11.2% in 1995 when all tariffs were eliminated.
Brazil’s tariffs could be capturing some omitted industry-level -time-varying variable. I address this problem in two ways. First, I include in the differenced equations 2-digit-ISIC-industry dummies that account for unobserved industry trends at broad sector levels like “Manufacture of food products and beverages” (ISIC 15) or “Manufacture of chemicals” (ISIC 24). As tariffs vary at the 4-digit-level this means that I am comparing manufacturers of dairy products (ISIC 1520) to macaroni producers (ISIC 1544), but not to manufacturers of pharmaceuticals (ISIC 2423) that are instead compared to producers of fertilizers (ISIC 2412). Second, as there can still be important differences between producers of pharmaceuticals and of fertilizers, I include 4-digit-ISIC-level controls for the industry characteristics that are likely to determine tariffs: the elasticity of demand, capital and skill intensity. These industry characteristics are measured with U.S. data to avoid endogeneity problems.

An additional issue concerning the use of Brazil’s tariffs to measure the effect of expanded export opportunities on entry in the export market and technology upgrading is that they might be correlated with changes in Argentina’s tariffs during this period, as long as the structure of protection was similar between the two countries in 1992. To address this concern I control for the change in Argentina’s tariffs with respect to the world in the period 1992-1996, and alternatively for the change in Argentina’s tariffs with respect to Brazil. I control both for final goods tariffs and intermediate inputs tariffs.

**Heterogeneous Effects**

The sorting pattern of firms described in Figure 2 implies that the reduction in Brazil’s tariffs should induce entry in the export market and technology upgrading for firms in the middle range.

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33 An important point to note is that as Argentina’s m.f.n tariffs with the rest of the world in 1992 were the basis for MERCOSUR tariff reductions, it is hard to distinguish the effect of the reduction of tariffs with respect to Brazil from changes of tariffs with respect to the rest of the world.
of the productivity distribution. In particular, the model predicts that the reduction in tariffs would induce firms in the middle range of the productivity distribution to enter the export market, but should not affect firms in the lower and upper ranges of the distribution. Similarly, the reduction in tariffs should only induce firms in the middle range of the productivity distribution to upgrade technology. To study these heterogeneous effects, I use firm size relative to the 4-digit-ISIC industry mean in 1992 as a proxy for initial productivity and divide firms into quartiles. Then, I analyze the effects of the reduction in Brazil’s tariffs on each quartile of the firm size distribution.

Next I present the estimation of the effect tariff changes on entry in the export market and later the estimation for technology upgrading.

C. Entry in the Export Market

I estimate a linearized version of the entry in the export market choice described by equation (7). This linearization does not respect functional form thus estimation only attempts to recover the signs of the partial derivative of interest and to assess the economic significance of the estimated coefficients. To simplify the exposition, I first describe estimation of the average effect of a reduction in Brazil’s tariffs on entry in the export market for all firms, and later analyze how this effect varies for firms in different quartiles of the size distribution.

I empirically analyze the entry in the export market decision using an index model:

\[
\text{EXP}_{ijt} = \begin{cases} 
1 & \text{if } \beta_t \tau^t_{jt} + \alpha_{st} + \mu_i + \epsilon_{ijt} > 0 \\
0 & \text{otherwise}
\end{cases}
\]

where \( j \) indexes 4-digit-ISIC industries; \( s \) indexes 2-digit-ISIC industries; \( t \) indexes time, that is the years 1992 and 1996; \( i \) indexes firms; \( \text{EXP}_{ijt} \) is a dummy variable that takes the value of 1 if
the firm exported in year $t$; $\tau^i_t$ are Brazil’s tariffs that vary across 4-digit-ISIC industries and time; $\mu_i$ are plant fixed effects that capture unobserved constant plant heterogeneity ($\phi$), constant sector characteristics that affect the sector exporting cutoffs in the model ($\sigma, k, f, f_e$) and also some other sector characteristics that although not included in the model might affect the exporting cutoffs (like factor intensity); $\alpha_{ij}$ are 2-digit-ISIC industry dummies that capture variation across time in sector characteristics.

Equation (10) with plant fixed effects can’t be consistently estimated by probit (incidental parameters problem). Then I estimate it using the linear probability model:

$$EXP_{ijst} = \beta \cdot \tau^i_t + \alpha_{ij} + \mu_i + \epsilon_{ijst}$$

In this case, first differencing eliminates time-invariant plant and sector heterogeneity:

$$\Delta EXP_{ijst} = \beta \cdot \Delta \tau^i + \Delta \alpha_{ij} + \Delta \epsilon_{ijst}$$

Estimation of equation (11) by OLS is reported in the first column of Panel A of Table 3, where the reported standard errors are clustered at the 4-digit-ISIC industry level. The coefficient in the change in Brazil’s tariffs ($\beta \cdot \tau^i$) is negative (-0.421) and significant ($t = -5.01$). The estimated coefficient implies that the average reduction in Brazil’s tariffs (24 percentage points) increases the probability of entry in the export market by 10 percentage points. Columns 2 to 8 assess the robustness of the baseline results to inclusion of controls, as described by the following equation:

$$\Delta EXP_{ijst} = \beta \cdot \Delta \tau^i + \beta \cdot \Delta \alpha_{ij} + \beta \cdot \Delta \epsilon_{ijst}$$

where $\Delta \tau^i$ denotes changes in Argentina’s import tariffs for outputs and inputs w.r.t. the world and Brazil; $z_{ij1992}$ are firm characteristics in the initial year (1992) like size measured by the

34 Bernard, Redding and Schott (2007) develop a two factor, two sector and two country model of trade with heterogeneous firms and show that the cutoff for entry in the export market is closer to the exit cutoff in comparative advantage industries.
number of workers, sales per worker and skill intensity; and \( c_j \) are 4-digit-ISIC industry characteristics like the elasticity of demand, skill and capital intensity in the U.S. Estimation of equation (12) is reported in columns (2) to (8) of Table 3, and although some of the firm and industry characteristics are highly significant, the coefficient on Brazil’s tariffs is not significantly affected by their inclusion. The coefficients in the regressions including all controls (columns 5 and 8) are -0.466 (t=-4.80) and -0.533 (t=-3.78) and imply that the average reduction in Brazil’s tariffs (24 percentage points) increases the probability of entry in the export market by 11 to 12 percentage points.

A potential problem of the specification in equation (12) is that if there are sunk exporting costs, current export status might depend on lagged export status,\(^{35}\) which in turn is likely to be negatively correlated with the initial level of Brazil’s tariffs. This problem can’t be solved by including lagged export status in the specification in first differences, as in that case export status in 1992 would be both part of the dependent variable and a regressor, thus necessarily correlated with the error term.\(^{36}\) Still, it is possible to estimate the equation in levels, including lagged export status as a regressor, as specified in the following equation:

\[
(13) \quad EXP_{ij1996} = \beta \tau_j + \delta EXP_{ij1992} + \alpha_x + \epsilon_{ij1996}
\]

Unlike the first-differenced specification, eq. (13) does not control for unobserved constant heterogeneity. Still, estimation of equation (13) is useful because first-difference and lagged-dependent-variable estimates have a bracketing property: if the first-difference specification in (12) is correct, then (13) will tend to underestimate the absolute value of \( \beta \), while if the

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\(^{35}\) Mark J. Roberts and Tybout (1997) and Bernard and Jensen (2004) find evidence of the existence of sunk exporting costs in Colombia and the U.S., respectively.

\(^{36}\) An alternative solution to this problem that permits to control both for unobserved individual heterogeneity and lagged dependent variables is to run a specification in first differences and use further lags of the dependent variable as instruments, as proposed by Manuel Arellano and Stephen Bond (1991). I can’t implement this solution because the panel I analyze only contains data for 1992 and 1996.
lagged-dependent-variable specification in (13) is correct, then (12) will tend to overestimate the absolute value $\beta_{\tau}$. This is because the initial level of Brazil’s tariffs is negatively correlated with export status in 1992. Panel B of Table 3 reports estimation of equation (13) where the estimated coefficient goes from -0.291 (t=4.09) in the baseline specification in Column 1 to -0.490 (t=3.40) in Column 9 where all controls are included. These estimates are 31% to 8% lower than the estimates in the first-differences specification, as expected. They are also less stable, possibly due to the omission of unobserved-time-invariant industry characteristics.

As a final check that the presence of sunk export costs is not creating a problem in the identification on the coefficient on Brazil’s tariffs I estimate equation (13) restricted to firms that were not exporters in 1992. Panel C of Table 3 reports the estimation of equation (13) by OLS. The coefficient on the change in Brazil’s tariffs is very similar to the one estimated in the first-difference specification and significant [-0.447 (t=-3.24) and -0.604 (t=-2.99) in columns 5 and 8 where all controls are included], implying that the average reduction in tariffs increases the probability of entering the export market by 11 to 14.5 percentage points.

A potential problem in the estimation of equation (13) restricted to non-exporters in 1992 is sample selection. The model predicts that in sectors where tariffs are higher the exporting cutoff is higher, thus it is likely that in sectors with high initial tariffs non exporters are more productive than in sectors with low initial tariffs, creating a positive correlation between Brazil’s tariffs in 1992 and unobserved productivity, thus biasing downwards the coefficient on the change in tariffs. A simple way to assess whether this is a problem is to look at the correlation of tariffs with firm characteristics that are correlated with unobserved productivity like size and

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37 For a discussion and a derivation of the biases in the first-difference and lagged-dependent-variable estimators see pages 243-247 in Joshua D. Angrist and Jorn-Steffen Pischke (2008).

38 Equation (13) can also be estimated by Probit as it does not contain firm fixed-effects. Probit estimation for both the full sample and the sample of non-exporters in 1992 produces very similar results as OLS. Tables reporting these estimations are available upon request.
sales per worker in the sub sample of non exporters in 1992, and both are very low (-0.033 and 0.013). In addition, when these firm characteristics are included in the regressions the coefficient does not change (see Panel C, columns 1 and 2), thus sample selection does not seem to play an important role.

[Insert Table 3]

Entry in the Export Market by Quartile of the Firm Size Distribution

The model predicts that the reduction in Brazil’s tariffs induces entry in the export market for firms in the middle range of the productivity distribution, but not for the least productive firms nor the most productive firms who would export even in the presence of high tariffs. More precisely, the prediction is that the reduction in tariffs induces entry for firms who were below the exporting threshold before liberalization, but above it afterwards. That is, those firms with productivity in the range $\phi^1 < \phi < \phi^0$ in Figure 2. To test this prediction, I estimate the effect of the change in Brazil’s tariffs on each quartile of the initial firm size distribution through the following equation:

$$\Delta \text{EXP}_{ijr} = \sum_{r=1}^{4} \beta_r^* \left( \Delta \tau^*_r \times Q^r_{ij} \right) + \sum_{r=2}^{4} \delta^r Q^r_{ij} + \Delta \alpha_i + \Delta \epsilon_{ijr}$$

where $r$ indexes each of the four quartiles of the size distribution and $Q^r_{ij}$ are dummy variables taking the value of 1 when firm $i$ belongs to quartile $r$. Estimation results are presented in column 1 Table 4. The effect of the reduction in Brazil’s tariffs on the probability to enter the export market is around 2 times larger in the 3rd quartile of the firm size distribution, where the point

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39 As a proxy for initial productivity, I use initial firm size in terms of (log) employment in primary school equivalents relative to the 4-digit-industry average, as detailed in Appendix D. Alternatively I used (log) domestic sales relative to the 4-digit-industry mean as a proxy for initial productivity, with similar but less precise results than the ones reported below. I prefer the employment measure because it reflects value-added better than sales, as long as there are differences in the level of vertical integration across firms.
estimate is -0.722 (t= 4.35). Column 4 presents estimation of the equation in levels including lagged export status as a control. The point estimates of β_c^τ are smaller, but the same pattern is observed: the estimate of β_c^τ is largest in the 3rd quartile and precisely estimated (β_c^3 =-0.541, t= -3.49). Column 7 reports estimation of β_c^τ in the sample of firms that were not exporters in 1992, with similar results as in the full sample (β_c^3 =-0.774, t= -3.20).

The effect of the reduction in tariffs on the rest of the quartiles is less precisely estimated. The coefficients are negative but not always statistically significant. Taken altogether the results suggest that some firms in the 1st, 2nd and 4th quartiles were induced to enter the export market by the reduction in Brazil’s tariffs. This is not inconsistent with the model, as size is not a perfect measure of productivity and the exporting cutoffs might differ across industries.

The point estimates of β_c^3 in the baseline specifications (columns 1, 4 and 7) imply that the 24 p.p. reduction in Brazil’s tariffs increases the probability to enter the export market by 19 to 13 p.p. for firms in the 3rd quartile of the size distribution. The finding that the reduction in Brazil’s tariffs had a smaller impact on entry for firms in the top quartile of the size distribution suggests that most of them were above the threshold before (or regardless of) liberalization. Similarly, the lower induced entry for firms in the first and second quartiles suggests that most of them were still below the threshold after liberalization. Thus, trade liberalization induced more entry in the export market for firms in the upper-middle range of the size distribution.

To assess the robustness of the baseline estimates of β_c^τ discussed above (Columns 1, 4 and 7 of Table 4) I perform a similar series of checks as in the estimation of average industry-level effects of the reduction in Brazil’s tariffs in Table 3. The remaining columns in Table 4 show that results are robust to the inclusion of changes in Argentina’s import tariffs (for both
output and inputs and w.r.t. the world and Brazil) and industry characteristics (capital, skill intensity and elasticity of demand).

[Insert Table 4]

**D. Technology Adoption Decision**

**Spending in Technology**

The technology adoption decision described in the model (equation 8) is binary. In the data, I observe a continuous measure of spending in technology and also some binary measures of product and process innovation. The technology spending measure has the advantage of having a panel structure that can be used to control for unobserved firm and industry characteristics but the disadvantage that only a sub-sample of firms has positive ST in 1992 and 1996. This sample is not representative for the smallest firms, while the binary measures of technology contained in the survey are available for a representative sample. I first analyze the ST measure and later I also discuss the binary innovation measures.

I first describe estimation of the average effect of a reduction in Brazil’s tariffs on spending in technology for all firms, and later analyze how this effect varies for firms in different quartiles of the size distribution. In the model, a firm is more likely to adopt technology the lower is the technology adoption threshold \( \phi^h \) in its sector [equation (8)], and the higher is its own productivity \( \phi \). Then the level of spending in technology can be described by:

\[
\log ST_{ijst} = \beta_e \tau_{j}^{i} + \beta_{e} \tau_{j}^{m} + \alpha_{st} + \mu_{i} + \epsilon_{ijst}
\]

where \( \tau^{m} \) denotes Argentina’s import tariffs, as adoption of new technologies depends on both export and domestic revenues. As the survey has information on ST for all the years in the period 1992-1996, equation (15) could be estimated in levels using the data for all the available years. The problem with this estimation strategy is that it would induce serial correlation in the
error terms, as the variation across time in Brazil’s tariffs is fully determined by their level in 1992. As a result, the standard error of the estimated coefficients would understate their standard deviation, as noted by Marianne Bertrand, Esther Duflo, and Sendhil Mullainathan (2004). Thus, instead of estimating equation (15) in levels for all the available years, I implement one of their proposed solutions. I collapse the data in two periods, one before (1992) and one after liberalization (1993-1996) and take first differences.\footnote{An alternative would be to only use the information in 1992 and 1996. I chose the first option to exploit all the available information, and also to minimize the number of observations with zero ST. The first alternative gives very similar results, although the standard errors are slightly bigger.} I thus estimate equation (15) in first differences:

\begin{equation}
\Delta \log ST_{ij} = \beta \tau_j^s + \beta \tau_j^m + \Delta \alpha_s + \Delta \epsilon_{ij} \tag{16}
\end{equation}

where the change in ST in the liberalization period is:

\[
\Delta \log ST_{ij} = \frac{1}{4} \sum_{t=1993}^{1996} \log ST_{ijt} - \log ST_{ij1992}.
\]

Estimation of equation (16) by OLS is reported in Table 5. The coefficient on the change in Brazil’s tariffs is negative and significant in all specifications. The estimated coefficient in the baseline specification in column 1, where only the change in Brazil’s tariffs is included as a regressor is -1.079 (t=3.08) and implies that the average reduction in Brazil’s tariffs (24 percentage points) induces an increase in technology spending of 0.24 log points. The estimated coefficient is not affected by the inclusion of firm-level controls (Column 2) nor by the change in Argentina’s output and input tariffs with respect to the world (Columns 3 to 5). Instead, the inclusion of the change in Argentina’s output tariffs with respect to Brazil (Column 6) increases the coefficient to -1.437 (t=-3.21), possibly because these are correlated with Brazil’s tariffs but had an effect of the opposite sign in technology adoption, although not statistically significant. Finally, the inclusion of the change in Argentina’s input tariffs with respect to Brazil (Column 7) does not affect the estimated coefficient.
A further question is whether the reduction in Brazil’s tariffs also increases the technology intensity of production, in the sense of increasing the ratio of spending in technology to labor. This is stronger evidence that firms are actually changing their production technology, instead of just expanding production by increasing the use of all factors proportionally. Table E.1 in Appendix E reports estimates of equation (16), replacing the growth in spending in technology by the growth in spending in technology per worker as the dependent variable. The estimates of $\beta_t \tau$ are very similar to the ones reported in Table 5.41

An important caveat in the interpretation of the results presented in this section is that equation (16) can only be estimated on a sub-sample of firms that have positive ST in 1992 and 1993-1996, 894 out of the total of 1380 firms in the panel. Firms reporting a positive level of spending in technology tend to be bigger: only 14% of them belong to the first size quartile, while 33% belong to the fourth, as reported in Table D.2 in Appendix D. Thus, results might not be representative for the smallest firms. Instead, the binary measures of technology that I use to construct the innovation indexes are available for a larger sub-sample of firms (1310 firms) that is representative in terms of size, as around 25% of firms in the sub-sample belong to each size quartile.

Binary Measures of Technology

In this section I analyze alternative measures of technology. I use a set of questions on improvements in products and production process to construct indexes for the fraction of questions in each category and overall that were answered positively by the firm.

41 Similar results are also obtained when the outcome variable is the ratio of spending in technology to sales.
Table 6 reports OLS estimates of equation (16), replacing the change in spending in technology by indexes of innovation as a dependent variable. The coefficient on the change in Brazil’s tariffs is negative and significant for all and each type of innovation. Consistent with the results presented in the previous section, the estimated coefficient is sensitive to the inclusion of the change in Argentina’s tariffs with respect to Brazil as a control, changing from -0.30 (t=3.35) to -0.40 (t=3.42) (columns 2 and 3). It is possible that this is due to the reduction in Argentina’s output tariffs having an effect of innovation of the opposite sign, although the estimate is only marginally significant (Column 3). The estimated coefficient in column 2 implies that the average reduction in Brazil’s tariffs induces an increase of 0.07 in the innovation index, which is 19% of the average innovation index (0.38). As the index is constructed as the fraction of yes/no questions about product and process innovation to which the firm gave a positive answer, the result can be interpreted as 19% increase in the fraction of questions about innovation answered positively by the firm. The effect of a reduction in Brazil’s tariffs is of similar magnitude when the innovation index is disaggregated in product and process innovations.

[Insert Table 6]

Technology Adoption by Quartile of the Firm Size Distribution

The model predicts that the reduction in Brazil’s tariffs induces technology adoption for firms in the middle range of the productivity distribution, but not for the least productive firms who do not export nor the most productive firms who already adopted the high technology. More precisely, the prediction is that the reduction in tariffs induces technology upgrading for firms who were below the technology adoption threshold before liberalization, but above afterwards. That is, those firms with productivity in the range $\phi^h_1 < \phi < \phi^h_0$ in Figure 2. To test this
prediction, I estimate the effect of the change in Brazil’s tariffs on each quartile of the initial firm size distribution through the following equation:

$$\Delta \log ST_{ij} = \sum_{r=1}^{4} \beta_{r}^\prime (\Delta \tau_{ij}^r \times Q_{ij}^r) + \sum_{r=2}^{4} \delta^r Q_{ij}^r + \beta_{r}^\prime \Delta \tau_{ij}^r + \Delta \alpha_r + \Delta \epsilon_{ij}$$

where \( r \) indexes each of the four quartiles of the size distribution and \( Q_{ij}^r \) are dummy variables taking the value of 1 when firm \( i \) belongs to quartile \( r \). Estimation results are presented in Table 7 for both spending in technology and the innovation index. The reduction in tariffs induces a statistically significant increase in spending in technology only in the third quartile of the size distribution, where the estimated coefficient is -2.106 (\( t=3.46 \)) (column 1). The point estimate is double the size than the estimated average effect for all firms reported in Table 5, and more than double the estimated coefficient for the other three quartiles of the size distribution. The results on the innovation index parallel the findings with the ST measure: the reduction in tariffs induces a statistically significant increase in innovation only in the third quartile of the size distribution, where the point estimate of \( \beta_{r}^\prime \) is -0.359 (\( t=2.70 \)), as reported in column 4.

The effect of the reduction in tariffs on the rest of the quartiles is less precisely estimated. The coefficients are negative but not statistically significant, thus it is possible that some firms in the 1\(^{st}\), 2\(^{nd}\) and 4\(^{th}\) quartiles were induced to upgrade technology by the reduction in Brazil’s tariffs. As mentioned above, this is not inconsistent with the model, as size is not a perfect measure of productivity and the technology adoption cutoffs might differ across industries.

The point estimate of \( \beta_{r}^3 \) implies that the 24 p.p. reduction in Brazil’s tariffs induces firms in the third quartile of the size distribution to increase their spending in technology an average of

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42 As noted above, I use initial firm size in terms of (log) employment in efficiency units relative to the 4-digit-industry average as a proxy for initial productivity. Alternatively I used (log) domestic sales relative to the 4-digit-industry mean as a proxy for initial productivity, with similar but less precise results than the ones reported below.
0.50 log points. The finding that firms in the top quartile of the size distribution did not increase ST in response to the reduction in tariffs suggests that they were above the threshold before, or regardless of, liberalization. Similarly, the lower and not statistically significant increase in spending in technology for firms in the second quartile suggests that they were still below the threshold after liberalization. Thus, trade liberalization induced technology upgrading for firms in the upper-middle range of the size distribution.

To assess the robustness of the baseline estimates of $\beta_r$, discussed above (Columns 1 and 4 of Table 7) I perform a similar series of checks as in the estimation of average industry-level effects of the reduction in Brazil’s tariffs in Tables 5 and 6. Columns 1-6 in Table 7 show that results are robust to the inclusion of changes in Argentina’s import tariffs (for both output and inputs and w.r.t. the world and Brazil) and industry characteristics (capital, skill intensity and elasticity of demand).

To assess whether firms increased the technology intensity of production, I estimate eq. 17 replacing the outcome of interest by spending in technology per worker. The results parallel the findings with the ST measure: the reduction in tariffs induces a statistically significant increase in spending in technology only in the third quartile of the size distribution, where the estimated coefficient is $-2.061 (t=3.65)$ (column 1, Table E.2 in Appendix E).

Finally, to assess whether the reduction in Brazil’s tariffs affected both product and process innovation, I estimate equation 17 separately for each type of innovation index. I obtain similar

43 As a further robustness check, interactions between changes in Argentina’s import tariffs and firm size quartiles were included as controls in the estimation of equation 17. Alternatively, a set of interactions between industry characteristics and firm size quartile dummies were also included as controls. The estimated coefficient of the effect of Brazil’s tariffs on ST and innovation in the 3rd quartile of the firm size distribution is not affected by the inclusion of these controls, and is always significant at 1% confidence level. Tables reporting these estimations are available upon request.
results as the ones reported above with the aggregate index, as reported in Columns 4-9 of Table E.2 in Appendix E.

\textit{E. Mechanism}

In this section I discuss how the evidence presented above relates to the mechanism emphasized in the theoretical model, namely that trade liberalization generates an increase in revenues for exporters making it profitable for them to adopt the high technology. Finally, I provide evidence that the reduction in Brazil’s tariffs increased export sales to Brazil, and the reduction in Argentina’s tariffs w.r.t. Brazil reduced domestic sales.

\textit{Technology Upgrading by Export Status}

The finding that the reduction in tariffs induces an increase in Spending in Technology on the 3\textsuperscript{rd} quartile of the firm size distribution is consistent with the theoretical prediction that only firms who are induced to cross a size threshold by the increase in export sales upgrade technology. To explore this issue further, I split the sample of firms in two on the basis of initial export status and show that the reduction in Brazil’s tariffs induced technology upgrading in both sub-samples. This implies that firms that were already exporting in 1992 are induced to upgrade technology by the reduction in Brazil’s tariffs, which is consistent with technology upgrading being driven by the increase in revenues. If technology upgrading was driven by the mere act of exporting, Brazil’s tariffs would impact technology spending only through their induced entry in the export market.

I estimate the effect of the reduction in Brazil’s tariffs on ST and the innovation indexes for two sub-samples of firms, the ones that did not export in 1992 and the ones that did. The
model predicts that both groups upgrade technology if the ordering of cutoffs is $\phi^i < \phi^h < \phi^i < \phi^h$, as depicted in Figure 2. Estimation results when the outcome of interest is ST are reported in Table 8, where Panel A reports the estimation the sub-sample of firms that did not export in 1992 and Panel B for the sub-sample of firms that exported in 1992. The coefficient is similar to the one estimated for the full sample, and significant in almost all specifications. These findings are consistent with the within-industry patterns in the data presented in Table 2, namely that both continuing exporters and new entrants in the export market increase ST faster than non exporters in the same 4-digit industry. Finally, Table 9 reports the estimation of the impact of Brazil’s tariffs when the outcome of interest are the product and process innovation indexes, with similar results in both sub-samples.

[Insert Table 8]

[Insert Table 9]

Exports to Brazil

In this section I report evidence using COMTRADE data on exports from Argentina to Brazil at the 4-digit-ISIC-industry level of aggregation that has information on export sales by destination. The analysis of export sales at the firm-level did not produce consistent results, possibly because the data combines all destinations. The main differences between the industry-level and firm-level data on exports are that industry-level data is reported by destination, reflects the universe of exports instead of a sample, and changes in export sales at the industry level not only capture the changes in sales of continuing exporters but also of new exporters.

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44 The estimated coefficients for the impact of the reduction in Brazil’s tariffs on the change in (log) export sales at the firm-level are not significant nor robust, in the sense that the point estimates vary between 1.429 and 0.16 depending on the combination of controls. In addition, they have the wrong sign. To check whether this instability is due to the fact that the distribution of the dependent variable is very disperse, I dropped the observations in the top and bottom deciles. The coefficients become all negative and stable between -0.471 and -0.252 but still not statistically significant.
Table 10 reports estimation of the effect of the reduction in Brazil’s tariffs on (log) exports to Brazil. As the number of observations is smaller than in the firm-level data set, I try to assess the robustness of the estimates by reporting results with and without 2-digit-ISIC industry dummies, and report both O.L.S. and I.V estimates. The I.V. estimates use Brazil’s tariffs in 1991 to instrument for Brazil’s tariffs in 1992 in an attempt to correct for measurement error in Brazil’s tariffs. As industry-level tariffs are computed as averages of product-level tariffs, zero or small trade in some products in a given year can produce inaccurate measures of industry-level tariffs. Panel A reports O.L.S estimates. The point estimate of the effect of Brazil’s tariffs on exports is negative, and statistically significant in all columns except in column 3 where controls for changes in Argentina’s tariffs w.r.t Brazil and 2-digit-ISIC industry dummies are included in the regression. Panel B reports I.V. estimates, where both the magnitude of the coefficient increases and standard errors fall, suggesting that measurement error in tariffs might produce attenuation bias in the O.L.S. results. Finally, Panel C reports the first stage of the I.V. estimates.

Overall, the industry-level data estimates suggest that the reduction in Brazil’s tariffs had a sizable impact on export sales: the 0.24 p.p. reduction in tariffs increased export sales by 0.68 to 0.84 log points, according to the O.L.S and I.V. baseline estimates reported in column 1, where 2-digit-ISIC-industry dummies are included.

[Insert Table 10]

**Domestic Sales**

The model predicts that domestic sales decline with tariff declines. The mapping of this prediction to the data is not straightforward, as the model considers a fully symmetric case where
changes in tariffs are the same for both countries. Thus, the model does not differentiate between Brazil’s and Argentina’s tariffs. The empirical evidence suggests that the reduction in Argentina’s tariffs w.r.t. Brazil reduced domestic sales, while the reduction in Brazil’s tariffs did not have a significant effect.

Table 11 reports estimation of the impact of the reduction in Argentina’s tariffs w.r.t. Brazil on domestic sales. The point estimate in the baseline specification in column 1 is 1.315 (t=2.41), and implies that the 13 p.p. reduction in Argentina’s tariffs reduced domestic sales by 0.17 log points. The point estimate is robust to the inclusion of controls for changes in Argentina’s input tariffs w.r.t. Brazil and industry characteristics (Columns 2 and 4), falls to 1.022 (t=1.89) when changes in Brazil’s tariffs are included in the regression (Column 5) and becomes insignificant when in addition controls for industry or firm characteristics are included (Columns 6 and 7). The evidence is thus not conclusive but suggestive that the reduction in Argentina’s tariffs w.r.t. Brazil reduced domestic sales.

Table 12 reports estimation of the effect of other changes in tariffs on domestic sales. Unlike the change in Argentina’s tariffs w.r.t. Brazil, changes in Argentina’s output and input tariffs w.r.t. the world did not have a statistically significant effect on domestic sales. This might be due to the fact that the changes in Argentina’s tariffs w.r.t the world were smaller, as unilateral trade liberalization took place before 1992. Finally, changes in Brazil’s tariffs w.r.t Argentina did not have a statistically significant effect on domestic sales.

[Insert Table 11]

[Insert Table 12]

45 The reason for considering the symmetric case is to obtain a closed-form solution for the model in general equilibrium, that is, allowing for the free entry of firms. This is important, as it highlights that trade liberalization has an impact on technology adoption only when not all firms export, as if all firms exported the increased revenues produced by tariff reductions would induce entry until revenues fall to their initial level, as indicated in the theory section.
IV. Concluding Remarks

The evidence reported in this paper suggests that expanded export opportunities can have a positive effect on firm performance. The evidence is consistent with falling trading partner’s tariffs increasing revenues for exporters and making adoption of new technologies profitable for more firms. The finding that falling trading partner’s tariffs induce firms to take actions that can increase their productivity suggests that the cross-sectional differences between exporters and non-exporters are not completely explained by selection of the most productive firms into the export market, but are partly induced by participation in export markets. Therefore, trade policies oriented to facilitate access to foreign markets, like multilateral trade liberalizations, can have a positive effect on firm-level performance.
References


### Table 1: Brazil and Argentina’s m.f.n Tariffs

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<th>Average</th>
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<th>Maximum</th>
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Note: Industries refer to 4-digit-ISIC industries with available tariff data.

### Table 2: Differences between exporters and non exporters

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<td>[1.034]***</td>
<td>[1.104]</td>
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</table>

Note: Robust Standard Errors in Brackets. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Exporter premia are estimated from a regression of the form: \( \ln Y_{ij} = \alpha_0 + \alpha_1 NE_{ij} + \alpha_2 EE_{ij} + \alpha_3 EN_{ij} + I_j + \epsilon_{ij} \) where \( i \) indexes firms, \( j \) indexes 4-digit-ISIC industries; \( NE \) are new exporters (231 firms), \( EE \) are continuing exporters (556 firms), \( EN \) are firms that exported in 1992 but didn’t in 1996 (27 firms) and the reference category relative to which differences are estimated is non exporters (566 firms); \( I \) are industry dummies, and \( Y \) is the firm characteristic for which the differences are estimated.
Table 3: Entry in the Export Market

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<td>1342</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.03</td>
<td>0.04</td>
<td>0.04</td>
<td>0.04</td>
<td>0.05</td>
<td>0.04</td>
<td>0.04</td>
</tr>
<tr>
<td>Panel B: Full Sample. Dependent variable: export status in 1996</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-0.291</td>
<td>-0.285</td>
<td>-0.278</td>
<td>-0.203</td>
<td>-0.323</td>
<td>-0.262</td>
<td>-0.281</td>
</tr>
<tr>
<td>[0.071]***</td>
<td>[0.077]***</td>
<td>[0.074]***</td>
<td>[0.084]***</td>
<td>[0.101]***</td>
<td>[0.101]***</td>
<td>[0.111]***</td>
<td>[0.144]***</td>
</tr>
<tr>
<td>Export status in 1992</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Outputs</td>
<td>0.642</td>
<td>0.543</td>
<td>0.543</td>
<td>0.544</td>
<td>0.546</td>
<td>0.542</td>
<td>0.545</td>
</tr>
<tr>
<td>Inputs</td>
<td>[0.023]***</td>
<td>[0.028]***</td>
<td>[0.028]***</td>
<td>[0.028]***</td>
<td>[0.029]***</td>
<td>[0.029]***</td>
<td>[0.028]***</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.46</td>
<td>0.50</td>
<td>0.50</td>
<td>0.50</td>
<td>0.50</td>
<td>0.50</td>
<td>0.50</td>
</tr>
<tr>
<td>Remaining controls and number of observations are the same as in the corresponding column in Panel A</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-0.411</td>
<td>-0.446</td>
<td>-0.457</td>
<td>-0.294</td>
<td>-0.447</td>
<td>-0.330</td>
<td>-0.357</td>
</tr>
<tr>
<td>[0.108]***</td>
<td>[0.124]***</td>
<td>[0.122]***</td>
<td>[0.123]***</td>
<td>[0.138]***</td>
<td>[0.150]***</td>
<td>[0.175]***</td>
<td>[0.202]***</td>
</tr>
<tr>
<td>Observations</td>
<td>797</td>
<td>797</td>
<td>797</td>
<td>781</td>
<td>781</td>
<td>797</td>
<td>781</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.04</td>
<td>0.16</td>
<td>0.16</td>
<td>0.16</td>
<td>0.17</td>
<td>0.16</td>
<td>0.16</td>
</tr>
<tr>
<td>Controls are the same as in the corresponding column in Panel A</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
</tr>
</tbody>
</table>

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Firm-level controls include employment measured in efficiency units, sales per worker and skill intensity, all measured in the initial year (1992). Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-SIC industry in the U.S.
Table 4: Entry in the Export Market by Quartile of the Firm Size Distribution

<table>
<thead>
<tr>
<th>Change in Brazil’s tariffs</th>
<th>Full Sample</th>
<th>Sample of non-exporters in 1992</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1) (2) (3)</td>
<td>(4) (5) (6) (7) (8) (9)</td>
</tr>
<tr>
<td>First Size Quartile</td>
<td>-0.331  -0.388  -0.442</td>
<td>-0.125  -0.159  -0.293</td>
</tr>
<tr>
<td>[0.185]**</td>
<td>[0.188]**  [0.198]**</td>
<td>[0.170]  [0.179]  [0.190]</td>
</tr>
<tr>
<td>Second Size Quartile</td>
<td>-0.327  -0.367  -0.412</td>
<td>-0.146  -0.170  -0.306</td>
</tr>
<tr>
<td></td>
<td>[0.146]**  [0.195]**  [0.219]**</td>
<td>[0.175]  [0.212]  [0.228]</td>
</tr>
<tr>
<td>Third Size Quartile</td>
<td>-0.722  -0.784  -0.832</td>
<td>-0.541  -0.576  -0.702</td>
</tr>
<tr>
<td></td>
<td>[0.166]*** [0.151]***  [0.203]***</td>
<td>[0.155]*** [0.152]***  [0.199]***</td>
</tr>
<tr>
<td>Fourth Size Quartile</td>
<td>-0.356  -0.429  -0.483</td>
<td>-0.286  -0.339  -0.474</td>
</tr>
<tr>
<td></td>
<td>[0.175]**  [0.179]**  [0.204]**</td>
<td>[0.119]*** [0.146]***  [0.162]***</td>
</tr>
</tbody>
</table>

Controls
Export Status in 1992
- Yes
- 0.553  0.558  0.557
- [0.027]*** [0.027]***  [0.028]***

Change in Arg.’s tariffs w.r.t. world
- Yes
- [0.027]***

Change in Arg.’s tariffs w.r.t. Brazil
- Yes
- Yes

Industry-level controls
- Yes
- Yes
- Yes

Firm-level controls
- Yes
- Yes
- Yes

2-digit-ISIC industry dummies
- Yes
- Yes
- Yes

Observations
- 1380  1348  1342
- 1380  1348  1342
- 797  781  781

R-squared
- 0.03  0.05  0.04
- 0.50  0.50  0.50
- 0.15  0.16  0.16

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Controls for changes in Argentina’s tariffs w.r.t. the world and Brazil include both output and input tariffs. Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S. Firm-level controls include dummies for the second, third and fourth quartile of the firm-size distribution in the initial year (1992).
**Table 5: Technology Adoption**  
Dependent variable: change in log (spending in technology)

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-1.079</td>
<td>-1.077</td>
<td>-1.051</td>
<td>-1.079</td>
<td>-0.939</td>
<td>-1.437</td>
<td>-1.435</td>
<td>-1.449</td>
</tr>
<tr>
<td></td>
<td>[0.350]**</td>
<td>[0.345]**</td>
<td>[0.325]**</td>
<td>[0.340]**</td>
<td>[0.383]**</td>
<td>[0.447]**</td>
<td>[0.483]**</td>
<td>[0.643]**</td>
</tr>
<tr>
<td>Change in Arg.’s tariffs w.r.t. world Outputs</td>
<td>0.556</td>
<td>0.599</td>
<td>0.629</td>
<td>2.051</td>
<td>2.254</td>
<td>2.538</td>
<td>[1.116]</td>
<td>[1.147]</td>
</tr>
<tr>
<td></td>
<td>[1.116]</td>
<td>[1.147]</td>
<td>[1.186]</td>
<td>[1.322]</td>
<td>[1.367]</td>
<td>[1.881]</td>
<td>[3.211]</td>
<td>[3.034]</td>
</tr>
<tr>
<td>Inputs</td>
<td>-0.762</td>
<td>-0.897</td>
<td>0.241</td>
<td>0.086</td>
<td>0.360</td>
<td>0.263</td>
<td>-0.584</td>
<td>0.187</td>
</tr>
<tr>
<td></td>
<td>[3.211]</td>
<td>[3.034]</td>
<td>[0.618]</td>
<td>[0.740]</td>
<td>[0.574]</td>
<td>[0.554]</td>
<td>[0.578]</td>
<td>[0.586]</td>
</tr>
</tbody>
</table>

Firm-level controls  Yes Yes Yes Yes Yes Yes Yes Yes  
Industry-level controls Yes Yes Yes Yes Yes Yes Yes Yes  
2-digit-ISIC industry dummies Yes Yes Yes Yes Yes Yes Yes Yes  
Observations 894 894 894 872 872 892 870 870  
R-squared 0.03 0.05 0.05 0.05 0.06 0.05 0.05 0.06  

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Firm-level controls include employment measured in efficiency units, sales per worker and skill intensity, all measured in the initial year (1992). Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S.

**Table 6: Product and Process Innovation**  
Dependent variable indicated in columns

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-0.236</td>
<td>-0.299</td>
<td>-0.400</td>
<td>-0.293</td>
<td>-0.346</td>
<td>-0.438</td>
<td>-0.183</td>
<td>-0.261</td>
<td>-0.357</td>
</tr>
<tr>
<td></td>
<td>[0.104]**</td>
<td>[0.090]**</td>
<td>[0.117]**</td>
<td>[0.116]**</td>
<td>[0.098]**</td>
<td>[0.116]**</td>
<td>[0.100]**</td>
<td>[0.093]**</td>
<td>[0.127]**</td>
</tr>
<tr>
<td>Change in Arg.’s tariffs w.r.t. world</td>
<td>-0.191</td>
<td>-0.184</td>
<td>0.241</td>
<td>0.086</td>
<td>0.360</td>
<td>0.263</td>
<td>-0.584</td>
<td>0.187</td>
<td>0.341</td>
</tr>
<tr>
<td></td>
<td>[0.271]</td>
<td>[0.284]</td>
<td>[0.554]</td>
<td>[0.578]</td>
<td>[0.586]</td>
<td>[0.308]*</td>
<td>[0.335]*</td>
<td>[0.326]</td>
<td>[0.326]</td>
</tr>
<tr>
<td>Inputs</td>
<td>0.530</td>
<td>0.626</td>
<td>0.530</td>
<td>0.626</td>
<td>0.530</td>
<td>0.626</td>
<td>0.530</td>
<td>0.626</td>
<td>0.530</td>
</tr>
<tr>
<td></td>
<td>[0.308]*</td>
<td>[0.335]*</td>
<td>[0.308]*</td>
<td>[0.335]*</td>
<td>[0.308]*</td>
<td>[0.335]*</td>
<td>[0.308]*</td>
<td>[0.335]*</td>
<td>[0.308]*</td>
</tr>
</tbody>
</table>

Industry-level controls  Yes Yes Yes Yes Yes Yes Yes Yes  
Firm-level controls  Yes Yes Yes Yes Yes Yes Yes Yes  
2-digit-ISIC industry dummies Yes Yes Yes Yes Yes Yes Yes Yes  
Observations 1301 1269 1263 1312 1280 1274 1319 1287 1281  
R-squared 0.24 0.25 0.26 0.22 0.24 0.24 0.22 0.23 0.23  

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Firm-level controls include employment measured in efficiency units, sales per worker and skill intensity, all measured in the initial year (1992). Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S.
Table 7: Technology Adoption by Quartile of the Firm Size Distribution

<table>
<thead>
<tr>
<th>Change in Brazil’s tariffs</th>
<th>Change in Spending in Technology 1996-1992</th>
<th>Product and Process Innovation</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)  (2) (3)</td>
<td>(4) (5) (6)</td>
</tr>
<tr>
<td>× First Size Quartile</td>
<td>-0.872  -0.725 -1.235</td>
<td>-0.041 -0.076 -0.165</td>
</tr>
<tr>
<td></td>
<td>[0.604] [0.570] [0.755]</td>
<td>[0.116] [0.113] [0.143]</td>
</tr>
<tr>
<td>× Second Size Quartile</td>
<td>-0.846  -0.662 -1.171</td>
<td>-0.199 -0.227 -0.326</td>
</tr>
<tr>
<td></td>
<td>[0.569] [0.629] [0.828]</td>
<td>[0.149] [0.145] [0.163]**</td>
</tr>
<tr>
<td>× Third Size Quartile</td>
<td>-2.106  -1.927 -2.424</td>
<td>-0.359 -0.403 -0.465</td>
</tr>
<tr>
<td></td>
<td>[0.609]*** [0.627]*** [0.886]***</td>
<td>[0.133]*** [0.146]*** [0.171]***</td>
</tr>
<tr>
<td>× Fourth Size Quartile</td>
<td>-0.372  -0.146 -0.648</td>
<td>-0.190 -0.229 -0.319</td>
</tr>
<tr>
<td></td>
<td>[0.534] [0.563] [0.773]</td>
<td>[0.130] [0.132]* [0.154]**</td>
</tr>
</tbody>
</table>

Controls

- Change in Arg.’s tariffs w.r.t. World
- Change in Arg.’s tariffs w.r.t. Brazil
- Industry-level controls
- 2-digit-ISIC industry dummies
- Observations: 894 872 870 1301 1269 1263
- R-squared: 0.05 0.06 0.06 0.20 0.20 0.20

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Controls for changes in Argentina’s tariffs w.r.t. the world and Brazil include both output and input tariffs. Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S. Firm-level controls include dummies for the second, third and fourth quartile of the firm-size distribution in the initial year (1992).
Table 8: Technology Adoption by initial Export Status
Dependent variable: change in log (spending in technology) 1996-1992

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Panel A: Sample of non-exporters in 1992</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-1.073</td>
<td>-1.256</td>
<td>-1.304</td>
<td>-1.101</td>
<td>-0.972</td>
<td>-1.586</td>
<td>-1.869</td>
<td>-1.788</td>
</tr>
<tr>
<td></td>
<td>[0.520]**</td>
<td>[0.513]**</td>
<td>[0.498]**</td>
<td>[0.435]**</td>
<td>[0.459]**</td>
<td>[0.682]**</td>
<td>[0.763]**</td>
<td>[0.947]*</td>
</tr>
<tr>
<td><strong>Panel B: Sample of exporters in 1992</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-1.116</td>
<td>-1.043</td>
<td>-0.979</td>
<td>-1.191</td>
<td>-1.052</td>
<td>-1.348</td>
<td>-1.015</td>
<td>-1.153</td>
</tr>
<tr>
<td></td>
<td>[0.382]***</td>
<td>[0.380]***</td>
<td>[0.396]**</td>
<td>[0.403]***</td>
<td>[0.520]**</td>
<td>[0.484]***</td>
<td>[0.555]*</td>
<td>[0.786]</td>
</tr>
</tbody>
</table>

**Controls**
- Change in Arg.’s tariffs w.r.t. world
  - Outputs: Yes, Yes, Yes
  - Inputs: Yes, Yes, Yes
- Change in Arg.’s tariffs w.r.t. Brazil
  - Outputs: Yes, Yes, Yes
  - Inputs: Yes, Yes, Yes
- Firm-level controls: Yes, Yes, Yes, Yes, Yes, Yes, Yes, Yes
- Industry-level controls: Yes, Yes, Yes, Yes, Yes, Yes, Yes, Yes
- 2-digit-ISIC industry dummies: Yes, Yes, Yes, Yes, Yes, Yes, Yes, Yes

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Number of observations in Panel A is 417 or 407 when input tariffs are included as controls. Number of observations in Panel B is 477 in columns 1 to 3, 465 in columns 4 and 6, 475 in column 6 and 463 in columns 7 and 8. Firm-level controls include employment measured in efficiency units, sales per worker and skill intensity, all measured in the initial year (1992). Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S.

Table 9: Product and Process Innovation by initial Export Status
Dependent variable indicated in columns

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Panel A: Sample of non-exporters in 1992</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-0.179</td>
<td>-0.279</td>
<td>-0.368</td>
<td>-0.220</td>
<td>-0.315</td>
<td>-0.403</td>
<td>-0.145</td>
<td>-0.252</td>
<td>-0.334</td>
</tr>
<tr>
<td></td>
<td>[0.117]</td>
<td>[0.098]***</td>
<td>[0.126]***</td>
<td>[0.126]*</td>
<td>[0.109]***</td>
<td>[0.126]***</td>
<td>[0.115]</td>
<td>[0.097]**</td>
<td>[0.138]**</td>
</tr>
<tr>
<td><strong>Panel B: Sample of exporters in 1992</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>-0.324</td>
<td>-0.331</td>
<td>-0.413</td>
<td>-0.403</td>
<td>-0.393</td>
<td>-0.455</td>
<td>-0.247</td>
<td>-0.277</td>
<td>-0.363</td>
</tr>
<tr>
<td></td>
<td>[0.130]**</td>
<td>[0.140]**</td>
<td>[0.184]**</td>
<td>[0.159]**</td>
<td>[0.151]**</td>
<td>[0.204]**</td>
<td>[0.122]**</td>
<td>[0.145]*</td>
<td>[0.181]**</td>
</tr>
</tbody>
</table>

**Controls**
- Change in Arg.’s tariffs w.r.t. world
  - Outputs: Yes, Yes, Yes
  - Inputs: Yes, Yes, Yes
- Change in Arg.’s tariffs w.r.t. Brazil
  - Outputs: Yes, Yes, Yes
  - Inputs: Yes, Yes, Yes
- Firm-level controls: Yes, Yes, Yes, Yes, Yes, Yes, Yes, Yes
- Industry-level controls: Yes, Yes, Yes, Yes, Yes, Yes, Yes, Yes
- 2-digit-ISIC industry dummies: Yes, Yes, Yes, Yes, Yes, Yes, Yes, Yes

Notes: standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Number of observations in Panel A is 741 in column 1, 725 columns 2, and 3, 747 in column 4, 731 in columns 5 and 6, 753 in column 7 and 737 in columns 8 and 9. Controls for changes in Argentina’s tariffs w.r.t. the world and Brazil include both output and input tariffs Firm-level controls include employment measured in efficiency units, sales per worker and skill intensity, all measured in the initial year (1992). Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S.
### Table 10: Export Sales to Brazil

**Dependent variable:** change in (log) export sales 1992-1996

<table>
<thead>
<tr>
<th>Panel A: OLS</th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Change in Brazil’s tariffs</strong></td>
<td>-2.836</td>
<td>-3.598</td>
<td>-2.402</td>
<td>-3.113</td>
<td>-2.621</td>
<td>-2.291</td>
</tr>
<tr>
<td>[1.560]*</td>
<td>[1.346]***</td>
<td>[1.622]</td>
<td>[1.325]**</td>
<td>[1.221]**</td>
<td>[1.325]*</td>
<td></td>
</tr>
<tr>
<td>R-squared</td>
<td>0.39</td>
<td>0.45</td>
<td>0.45</td>
<td>0.07</td>
<td>0.19</td>
<td>0.17</td>
</tr>
<tr>
<td><strong>Panel B: IV</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>Change in Brazil’s tariffs</strong></td>
<td>-3.513</td>
<td>-4.226</td>
<td>-3.129</td>
<td>-3.911</td>
<td>-3.326</td>
<td>-3.242</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.39</td>
<td>0.45</td>
<td>0.45</td>
<td>0.07</td>
<td>0.19</td>
<td>0.16</td>
</tr>
<tr>
<td><strong>Panel C: First Stage</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Brazil’s tariffs in 1991</td>
<td>-0.774</td>
<td>-0.764</td>
<td>-0.747</td>
<td>-0.763</td>
<td>-0.767</td>
<td>-0.744</td>
</tr>
<tr>
<td>[0.035]***</td>
<td>[0.040]***</td>
<td>[0.038]***</td>
<td>[0.026]***</td>
<td>[0.029]***</td>
<td>[0.032]***</td>
<td></td>
</tr>
<tr>
<td>R-squared</td>
<td>0.95</td>
<td>0.95</td>
<td>0.95</td>
<td>0.94</td>
<td>0.94</td>
<td>0.94</td>
</tr>
<tr>
<td>F-test on excluded instrument</td>
<td>480.12</td>
<td>366.19</td>
<td>378.54</td>
<td>867.08</td>
<td>685.09</td>
<td>548.69</td>
</tr>
</tbody>
</table>

**Controls**
- Change in Arg.’s tariffs w.r.t. world: Yes
- Change in Arg.’s tariffs w.r.t. Brazil: Yes
- Industry-level controls: Yes
- 2-digit-ISIC industry dummies: Yes
- Observations: 100
- R-squared: 0.39

**Note:** Robust standard errors in brackets. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Controls for changes in Argentina’s tariffs w.r.t. the world and Brazil include both output and input tariffs. Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S.

### Table 11: Domestic Sales, Effect of Argentina’s Output Tariff Reductions w.r.t Brazil

**Dependent variable:** change in log (domestic sales) 1996-1992

<table>
<thead>
<tr>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Change in Arg.’s tariffs w.r.t. Brazil</td>
<td>1.315</td>
<td>1.465</td>
<td>1.022</td>
<td>1.278</td>
<td>1.022</td>
<td>0.662</td>
</tr>
<tr>
<td>Outputs</td>
<td>[0.545]**</td>
<td>[0.573]**</td>
<td>[0.541]*</td>
<td>[0.545]**</td>
<td>[0.538]*</td>
<td>[0.585]</td>
</tr>
<tr>
<td>Inputs</td>
<td>-1.010</td>
<td>-0.591</td>
<td>-0.579</td>
<td>-0.579</td>
<td>-1.161</td>
<td>-1.047</td>
</tr>
<tr>
<td>[1.556]</td>
<td>[1.417]</td>
<td>[1.351]</td>
<td>[1.285]</td>
<td>[1.241]</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>0.191</td>
<td>0.306</td>
<td>0.438</td>
<td></td>
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<td></td>
</tr>
<tr>
<td>[0.256]</td>
<td>[0.211]</td>
<td>[0.271]</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Firm-level controls: Yes</td>
<td>Yes</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Industry-level controls: Yes</td>
<td>Yes</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2-digit-ISIC industry dummies: Yes</td>
<td>Yes</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Observations: 1371</td>
<td>1339</td>
<td>1339</td>
<td>1339</td>
<td>1371</td>
<td>1339</td>
<td>1339</td>
</tr>
<tr>
<td>R-squared: 0.06</td>
<td>0.06</td>
<td>0.08</td>
<td>0.06</td>
<td>0.06</td>
<td>0.08</td>
<td>0.06</td>
</tr>
</tbody>
</table>

**Notes:** standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%. Firm-level controls include employment measured in efficiency units and skill intensity, both measured in the initial year (1992). Industry-level controls include demand elasticity, skill intensity and capital intensity of the 4-digit-ISIC industry in the U.S.

### Table 12: Domestic Sales, Effect of other Tariff Reductions

**Dependent variable:** change in log (domestic sales) 1996-1992

<table>
<thead>
<tr>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Change in Brazil’s tariffs</td>
<td>0.368</td>
<td>0.579</td>
<td></td>
<td></td>
</tr>
<tr>
<td>[0.246]</td>
<td>[1.442]</td>
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<tr>
<td>Change in Arg.’s tariffs w.r.t. Brazil</td>
<td>0.579</td>
<td>-1.171</td>
<td>-1.425</td>
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</tr>
<tr>
<td>Outputs</td>
<td>[0.786]</td>
<td>[0.788]*</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Inputs</td>
<td>2.358</td>
<td>2.551</td>
<td></td>
<td></td>
</tr>
<tr>
<td>[1.592]</td>
<td>[1.635]</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>2-digit-ISIC industry dummies: Yes</td>
<td>Yes</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Observations: 1377</td>
<td>1345</td>
<td>1377</td>
<td>1345</td>
<td>1345</td>
</tr>
<tr>
<td>R-squared: 0.05</td>
<td>0.05</td>
<td>0.05</td>
<td>0.05</td>
<td>0.05</td>
</tr>
</tbody>
</table>

**Notes:** standard errors are clustered at the 4-digit-ISIC industry level. * indicates significant at 10%; ** significant at 5%; *** significant at 1%.