

What Goods Do Countries Trade? A Quantitative Exploration of Ricardo's Ideas

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The Ricardian model predicts that countries should produce and export relatively more in industries in which they are relatively more productive. Though one of the most celebrated insights in the theory of international trade, this prediction has received little attention in the empirical literature since the mid-1960s. The main reason behind this lack of popularity is the absence of clear theoretical foundations to guide the empirical analysis. Building on the seminal work of Eaton and Kortum (2002, “Technology, Geography, and Trade”, *Econometrica*, **70**, 1741–1779), we offer such foundations and use them to quantify the importance of Ricardian comparative advantage. In the process, we also provide a theoretically consistent alternative to Balassa's (1965, “An Empirical Demonstration of Classical Comparative Cost Theory”, *Review of Economics and Statistics*, **45**, 231–238) well-known index of “revealed comparative advantage”.

Key words: Ricardian model, Comparative advantage, Productivity, Commodity pattern of trade

JEL Codes: F10, F11

1. INTRODUCTION

The Ricardian model predicts that countries should produce and export relatively more in industries in which they are relatively more productive. Though one of the most celebrated insights in the theory of international trade, this prediction has received little attention in the empirical literature since the mid-1960s. The main reason behind this lack of popularity is not the existence of strong beliefs regarding the (un)importance of technological considerations. It derives instead from the absence of any clear theoretical foundations to guide the empirical analysis (see, e.g. Bhagwati, 1964; Deardorff, 1984; Leamer and Levinsohn, 1995). Building on the seminal work of Eaton and Kortum (2002), we develop such foundations and use them to quantify the importance of Ricardian comparative advantage at the industry level. In the process, we also provide a theoretically consistent alternative to Balassa's (1965) well-known index of “revealed comparative advantage”.

The tight connection between theory and empirics offered by our paper has two crucial advantages over the previous literature (see, e.g. MacDougall 1951, Stern 1962, Balassa 1963, or more recently, Golub and Hsieh 2000). First, when estimating the importance of Ricardian

comparative advantage in the cross section, we do not have to rely on *ad hoc* measures of export performance such as total exports to the rest of the world, total exports to third markets, or bilateral net exports. Our theory tells us exactly what the dependent variable in our cross-sectional regressions ought to be: log of exports, disaggregated by exporting and importing countries, differenced across industries and exporters, and adjusted for differences in levels of “openness” to account for trade-driven selection. Using trade and productivity data, we can therefore offer the first theoretically consistent Ricardian test. Our preferred estimate implies that, *ceteris paribus*, the elasticity of (adjusted) bilateral exports with respect to observed productivity is positive, as our Ricardian model predicts, and equal to 6.53.

Second, our clear theoretical foundations allow us to do general equilibrium counterfactual analysis. In order to quantify further the importance of Ricardian comparative advantage, we ask: If, for any pair of exporters, there were no relative productivity differences across industries, what would be the consequences for aggregate trade flows and welfare? According to our estimates, the removal of Ricardian comparative advantage at the industry level would only lead, on average, to a 5.3% decrease in the total gains from trade.¹

Section 2 describes our theoretical framework. We consider an economy with multiple countries, multiple industries, and one factor of production, labour. Up to this point, this is a standard Ricardian model. We generalize this model by allowing for intra-industry heterogeneity à la Eaton and Kortum (2002). Formally, we assume that each good is available in many varieties and that labour productivity differs across varieties. The key feature of this model is that labour productivity may be separated into (i) a deterministic component, which is country and industry specific, and (ii) a stochastic component, randomly drawn across countries, industries, and varieties. The former, which we refer to as “fundamental productivity”, captures factors such as climate, infrastructure, and institutions that affect the productivity of all producers in a given country and industry. The latter, by contrast, reflects idiosyncratic differences in technological know-how across varieties.

Section 3 derives our theoretical predictions, relates them to the previous Ricardian literature, and investigates their robustness. The first set of predictions are cross-sectional in nature, and describe how productivity differences affect trade patterns across countries and industries in any trading equilibrium. A key result that emerges is a parsimonious method for correcting for the trade-driven selection that creates an endogenous wedge between fundamental and observed productivity. The second set of predictions concern general equilibrium responses to novel and natural counterfactual scenarios that explore by how much aggregate trade flows and welfare would change if, for any pair of exporters, there were no fundamental relative productivity differences across industries.

Section 4 tests our cross-sectional predictions using the best available data on internationally comparable productivity and trade flows across countries and industries. In line with our Ricardian model, our measures of relative productivity are computed using relative producer prices from the Groningen Growth and Development Centre (GGDC) Productivity Level Database. Our analysis illustrates how tests of the Ricardian model, which have long been perceived as hopelessly *ad hoc*, can be performed in a theoretically consistent manner. This procedure allows us to estimate the extent of intra-industry heterogeneity, “ θ ”, which is the key structural parameter governing the relationship between productivity and exports in this Ricardian world. After adjusting for differences in openness across exporting countries and industries, which turns out to be empirically important, our preferred estimate of θ is equal to 6.53. This is in line with

1. This quantification, of course, depends on the level of aggregation at which an “industry” is defined. Throughout this paper, industries will be defined as the lowest level of disaggregation for which we have productivity data that are comparable across countries and industries, namely, the two-digit International Standard Industrial Classification (ISIC) level.

previous estimates of θ obtained using different methodologies (see, *e.g.* Eaton and Kortum, 2002; Bernard *et al.*, 2003; Donaldson, 2008; Simonovska and Waugh, 2009).

Section 5 presents our counterfactual results. In order to minimize measurement error, we first demonstrate how to use our model to compute revealed measures of productivity. While not the main focus of our paper, this approach provides a theoretically consistent alternative to Balassa's (1965) well-known index of "revealed comparative advantage". Using these productivity measures, we then investigate the consequences for aggregate trade flows and welfare of removing relative productivity differences across countries and industries. According to our estimates, the removal of Ricardian comparative advantage at the industry level would only lead, on average across countries, to a 5.3% decrease in the total gains from trade. We conclude Section 5 by exploring how the magnitude of this estimate is related to two important features of the data: heterogeneous preferences and heterogeneous trade costs. Both of these aspects of heterogeneity tend to offset the purely productivity-driven heterogeneity that is at the heart of the Ricardian model.

Our paper is related both to previous empirical tests of the Ricardian model—MacDougall (1951), Stern (1962), Balassa (1963), and Golub and Hsieh (2000)—and to a much more recent but rapidly growing literature based on multisector extensions of the Eaton and Kortum (2002) model—Shikher (2004, 2008), Costinot (2005), Chor (2008), Donaldson (2008), Caliendo and Parro (2009), Kerr (2009), Burstein and Vogel (2010), and Levchenko and Zhang (2011). Relative to this literature, our first contribution is to show how micro-foundations à la Eaton and Kortum (2002) can be used to contrast the cross-sectional predictions of the Ricardian model with the data. Using our model, we estimate the impact of observed productivity differences on the pattern of trade across countries and industries without having to rely on *ad hoc* measures of export performance, bilateral comparisons inspired by a two-country model, or unclear orthogonality conditions.² Our second contribution is to show how estimates obtained from these regressions can be used to measure, in a well-defined way, the welfare impact of Ricardian comparative advantage at the industry level.³ Put together, these cross-sectional and counterfactual results provide a complete exploration of the quantitative implications of Ricardo's ideas.

Our analysis is also related, though less closely, to a large empirical literature investigating the role of various sources of comparative advantage (see, *e.g.* Harrigan, 1997; Beck, 2003; Romalis, 2004; Yeaple and Golub, 2007; Nunn, 2007; Manova, 2008; Morrow, 2010). Like the previous papers, we analyse how differences in costs of production may affect the pattern of trade across countries and industries. However, our main focus is not to assess which particular channel—*e.g.* institutions, infrastructure, or factor endowments—contributes more to the cross-sectional variation in costs of production, and in turn, bilateral trade flows. Like in previous Ricardian tests, we start by interpreting the cross-sectional variation in costs of production as the result of fundamental productivity differences. Conditional on this interpretation, we then investigate how much of the cross-sectional variation in trade flows and the overall gains from trade can be explained by these differences. In other words, the goal of the present paper is not to investigate what causes observed productivity differences or whether such differences may

2. It is worth emphasizing that, our theoretical foundations notwithstanding, our cross-sectional results are much closer, in spirit, to those of MacDougall (1951), Stern (1962), Balassa (1963), and Golub and Hsieh (2000) than to those of Eaton and Kortum (2002). The focus of our analysis is the commodity pattern of trade, not its total volume. Compared to Eaton and Kortum (2002), our empirical exercise therefore requires independent measures of productivity as well as trade data disaggregated by industry.

3. In this regard, the present analysis is most closely related to Chor (2008) who compares the impact of various sources of comparative advantage on welfare, albeit not in a fully structural way.

be accounted for by differences in factor endowments. Our goal simply is to ask: Seen through the lens of the Ricardian model, how important are productivity differences across countries and industries?

2. THEORETICAL FRAMEWORK

We consider a world economy comprising $i = 1, \dots, I$ countries and one factor of production, labour. There are $k = 1, \dots, K$ industries or goods and constant returns to scale in the production of each good. Labour is perfectly mobile across industries and immobile across countries. We denote by L_i and w_i the number of workers and the wage in country i , respectively. Up to this point, this is a standard Ricardian model. We generalize this model by allowing for intra-industry heterogeneity in labour productivity.

Technology. Each good k may come in an infinite number of varieties indexed by $\omega \in \Omega \equiv \{1, \dots, +\infty\}$.⁴ We denote by $z_i^k(\omega)$ the number of units of the ω th variety of good k that can be produced with one unit of labour in country i . In line with Eaton and Kortum (2002), we assume the following:

A1. For all countries i , goods k , and their varieties ω , $z_i^k(\omega)$ is a random variable drawn independently for each triplet (i, k, ω) from a Fréchet distribution $F_i^k(\cdot)$ such that

$$F_i^k(z) = \exp[-(z/z_i^k)^{-\theta}], \quad \text{for all } z \geq 0, \quad (1)$$

where $z_i^k > 0$ and $\theta > 1$.

According to Assumption A1, technological differences across countries and industries only depend on two parameters, z_i^k and θ . We refer to z_i^k as the *fundamental productivity* of country i in industry k . It aims to capture factors such as climate, infrastructure, and institutions that affect the productivity of *all* producers in a given country and industry. For each industry k , cross-country variation in z_i^k pins down the cross-country variation in relative labour productivity at the core of the standard Ricardian model. Formally, equation (1) implies that $z_i^k/z_{i'}^k = E[z_i^k(\omega)]/E[z_{i'}^k(\omega)]$, for any i, i' , and k .

By contrast, the second parameter, θ , measures *intra-industry heterogeneity*. It aims to reflect the scope for idiosyncratic differences in technological know-how across varieties, which we assume to be the same in all countries and industries.⁵ In this model, θ parameterizes the impact of changes in fundamental productivity levels, z_i^k , on aggregate trade flows. Estimating θ will be one key focus of our quantitative exploration of the Ricardian model. Note that the assumption that θ is common across industries is not innocuous. It will maintain the tight relationship between fundamental productivity and comparative advantage at the core of the standard Ricardian model by implicitly ruling out the possibility that differences in wages across countries may interact with differences in intra-industry heterogeneity across sectors to determine the pattern of commodity trade.⁶

4. Alternatively, we could have assumed the existence of a continuum of varieties. By assuming that the number of varieties is infinite, but countable, we avoid the technical difficulties of invoking the law of large numbers with a continuum of i.i.d. variables (see, e.g. Al-Najjar, 2004). Nothing substantial hinges on this particular modelling choice. More economically substantial is the implicit assumption that the number of varieties per industry—unlike in a monopolistically competitive model with free entry—is exogenously given.

5. Using a unique data set that documents physical output (rather than just revenue) at the plant level in the U.S., Foster, Haltiwanger and Syverson (2008) document significant intra-industry heterogeneity in plant-level physical productivity.

6. Fieler (2010) discusses the quantitative implications of these interaction effects for total trade volumes.

Trade costs. Trade frictions take the standard “iceberg” form. Formally, we assume the following:

A2. For each unit of good k shipped from country i to country j , only $1/d_{ij}^k \leq 1$ units arrive, with d_{ij}^k such that (i) $d_{ii}^k = 1$ and (ii) $d_{il}^k \leq d_{ij}^k \cdot d_{jl}^k$ for any third country l .

The second part of Assumption A2 simply rules out cross-country arbitrage opportunities.

Market structure. Markets are assumed to be perfectly competitive. Together with constant returns to scale in production, perfect competition implies the following:

A3. In any country j , the price $p_j^k(\omega)$ paid by buyers of variety ω of good k is

$$p_j^k(\omega) = \min_{1 \leq i \leq I} [c_{ij}^k(\omega)], \quad (2)$$

where $c_{ij}^k(\omega) = (d_{ij}^k \cdot w_i)/z_i^k(\omega)$ is the cost of producing and delivering one unit of this variety from country i to country j .

For each variety ω of good k , buyers in country j are “shopping around the world” for the best price available. In what follows, we let $c_{ij}^k = (d_{ij}^k \cdot w_i)/z_i^k > 0$.

Preferences. In each country, there is a representative consumer with a two-level utility function. The upper tier utility function is Cobb–Douglas, while the lower tier is constant elasticity of substitution (CES).⁷ Accordingly, expenditures are such that the following holds true:

A4. In any country j , total expenditure on variety ω of good k is

$$x_j^k(\omega) = [p_j^k(\omega)/p_j^k]^{1-\sigma_j^k} \cdot \alpha_j^k w_j L_j, \quad (3)$$

where $0 \leq \alpha_j^k \leq 1$, $\sigma_j^k < 1 + \theta$, and $p_j^k \equiv [\sum_{\omega' \in \Omega} p_j^k(\omega')^{1-\sigma_j^k}]^{1/(1-\sigma_j^k)}$.

The above expenditure function is a standard feature of models of intra-industry trade in the “new trade” literature (see, e.g. Helpman and Krugman, 1985). The first preference parameter, α_j^k , measures the share of expenditure on varieties from industry k in country j , whereas the second, σ_j^k , is the elasticity of substitution between varieties. The restriction $\sigma_j^k < 1 + \theta$ is a technical assumption that guarantees the existence of a well-defined CES price index, p_j^k , in country j and industry k . It is worth emphasizing that demand conditions may vary across countries and industries: α_j^k and σ_j^k are functions of j and k . For future reference, we denote by $p_j \equiv \prod_{k=1}^K (p_j^k)^{\alpha_j^k}$ the consumer price index in country j .

Trade balance. We denote by $x_{ij}^k \equiv \sum_{\omega \in \Omega_{ij}^k} x_j^k(\omega)$ the value of total exports from country i to country j in industry k , where $\Omega_{ij}^k \equiv \{\omega \in \Omega | c_{ij}^k(\omega) = \min_{1 \leq i' \leq I} c_{i'j}^k(\omega)\}$ is the set of varieties exported by country i to country j in industry k . Similarly, we denote by $\pi_{ij}^k \equiv x_{ij}^k / \sum_{i'=1}^I x_{i'j}^k$ the share of exports from country i in country j and industry k . Our final assumption is the following:

A5. For any country i , trade is balanced

$$\sum_{j=1}^I \sum_{k=1}^K \pi_{ij}^k \alpha_j^k \gamma_j = \gamma_i, \quad (4)$$

where $\gamma_i \equiv w_i L_i / \sum_{i'=1}^I w_{i'} L_{i'}$ is the share of country i in world income.

7. While the Cobb–Douglas assumption in A4 could be dispensed with in favour of any upper tier utility function for our cross-sectional results, it will play a non-trivial role in our counterfactual analysis.

Conditional on aggregate trade flows, Assumption A5 pins down relative wages, $w_i/w_{i'}$, around the world. This completes our description of a trade equilibrium in this economy.

3. THEORETICAL PREDICTIONS

Using the previous theoretical framework, Assumptions A1–A5, we derive two types of predictions. First, we show how differences in labour productivity across countries and industries affect the pattern of trade *in a given equilibrium*. These cross-sectional predictions will form the basis of our first empirical exercises (in Section 4), the goal of which is to estimate the key model parameter θ which governs the relationship between productivity and exports in this Ricardian world. Second, we demonstrate how changes in labour productivity would affect trade and welfare *across equilibria*. These counterfactual predictions will form the basis of our final empirical exercise (in Section 5), in which we explore the importance of inter-industry Ricardian forces for generating gains from trade around the world.

3.1. Productivity and trade: a first look

Before deriving our new cross-sectional predictions, we first describe the impact of fundamental productivity and trade costs on bilateral exports at the industry level. By themselves, such theoretical results are of limited use for empirical work because fundamental productivity levels cannot be observed: doing so would require data on products that have been driven out of domestic production by the forces of trade. This intermediate step, however, will allow us to compare our approach to the previous theoretical literature on the Ricardian model in a transparent manner. In the next subsection, we will then highlight how trade-driven selection drives a wedge between fundamental productivity and observed productivity and how empirical work can nevertheless proceed in the face of this wedge.

Lemma 1. *Suppose that Assumptions A1–A4 hold. Then for any importer, j , any pair of exporters, i and i' , and any pair of goods, k and k' ,*

$$\ln\left(\frac{x_{ij}^k x_{i'j}^{k'}}{x_{ij}^{k'} x_{i'j}^k}\right) = \theta \ln\left(\frac{z_i^k z_{i'}^{k'}}{z_i^{k'} z_{i'}^k}\right) - \theta \ln\left(\frac{d_{ij}^k d_{i'j}^{k'}}{d_{ij}^{k'} d_{i'j}^k}\right). \quad (5)$$

The formal proof of Lemma 1 as well as all subsequent proofs can be found in the online appendix. The basic idea is simple. Under Assumptions A1–A4, bilateral exports from country i to country j in sector k can be expressed as

$$x_{ij}^k = \frac{(w_i d_{ij}^k / z_i^k)^{-\theta}}{\sum_{i'=1}^I (w_{i'} d_{i'j}^k / z_{i'}^k)^{-\theta}} \cdot \alpha_j^k w_j L_j. \quad (6)$$

To go from this expression to equation (5), we then use a simple “difference-in-difference” strategy. The first (log-)difference, $x_{ij}^k/x_{ij}^{k'}$, allows us to control for differences in wages, w_i , across exporters, as well as differences in income, $w_j L_j$, across importers. The second (log-) difference, $(x_{ij}^k/x_{ij}^{k'})/(x_{i'j}^k/x_{i'j}^{k'})$, allows us to control for differences in the share of expenditure, α_j^k , across sectors. Equation (5) follows.

At this point, our theoretical framework imposes very little structure on the variation of trade costs. To better relate Lemma 1 to previous results in the literature, it is useful to consider briefly the special case in which

$$d_{ij}^k = d_{ij} \cdot d_j^k \quad \text{for all } k \text{ and } i \neq j. \quad (7)$$

This restriction is trivially satisfied, *e.g.* in [Dornbusch, Fischer and Samuelson \(1977\)](#) who assume that trade costs are symmetric across countries and identical across goods, $d_{ij}^k = d_{ji}^k \equiv d$. In equation (7), the first parameter d_{ij} measures the trade barriers which are specific to countries i and j such as physical distance, existence of colonial ties, use of a common language, or participation in a monetary union. The second parameter d_j^k measures the policy barriers imposed by country j on good k , such as import tariffs and standards, which, in line with “the most-favoured-nation” clause of the World Trade Organization, may not vary by country of origin.⁸

In this special case, since $d_{ij}^k d_{i'j}^{k'}/d_{ij}^{k'} d_{i'j}^k = 1$ for all $i, i' \neq j$, Lemma 1 directly implies the following:

Corollary 1. *Suppose that Assumptions A1–A4 and equation (7) hold. Then for any importer, j , and any pair of exporters, $i, i' \neq j$, the ranking of relative fundamental productivity determines the ranking of relative exports:*

$$\frac{z_i^1}{z_{i'}^1} \leq \dots \leq \frac{z_i^K}{z_{i'}^K} \Leftrightarrow \frac{x_{ij}^1}{x_{i'j}^1} \leq \dots \leq \frac{x_{ij}^K}{x_{i'j}^K}.$$

Note that, without loss of generality, we can always fix a pair of countries, i_1 and i_2 , and index the K goods so that

$$\frac{z_{i_1}^1}{z_{i_2}^1} \leq \dots \leq \frac{z_{i_1}^K}{z_{i_2}^K}. \quad (8)$$

Ranking (8) is at the heart of the standard two-country Ricardian model (see, *e.g.* [Dornbusch, Fischer and Samuelson, 1977](#)). When there is no intra-industry heterogeneity, ranking (8) merely states that country i_1 has a comparative advantage in (all varieties of) the high k goods. If there only are two countries, the pattern of trade follows: i_1 produces and exports the high k goods, while i_2 produces and exports the low k goods. If there are more than two countries, however, the pattern of pairwise comparative advantage no longer determines the pattern of trade. In this case, the standard Ricardian model loses most of its intuitive content (see, *e.g.* [Jones, 1961](#); [Wilson, 1980](#)).

When there are stochastic productivity differences within each industry, it is easy to check that Assumption A1 and ranking (8) further imply

$$\frac{z_{i_1}^1(\omega)}{z_{i_2}^1(\omega)} \preceq \dots \preceq \frac{z_{i_1}^K(\omega)}{z_{i_2}^K(\omega)}, \quad (9)$$

where \preceq denotes the first-order stochastic dominance order among distributions. In other words, ranking (9) is just a stochastic—hence weaker—version of the ordering of labour productivity z_i^k , which is at the heart of the Ricardian theory. Like its deterministic counterpart in equation (8), ranking (9) captures the idea that country i_1 is relatively better at producing the high k goods. But whatever k is, country i_2 may still have lower labour requirements on some of its varieties.⁹

8. In practice, equation (7) may be violated because of preferential trade agreements or, more simply, because bilateral distance has a differential impact on goods of different weights (see, *e.g.* [Harrigan, 2005](#)). For this reason, we do not impose equation (7) throughout the remainder of this paper. The role of equation (7) here is purely to relate our results to previous theoretical work on the Ricardian model.

9. It should be clear, however, that ranking (9) *per se* does not imply the equivalence in Corollary 1. To derive this equivalence, we use equation (5) which relies on the particular functional forms in A1.

According to Corollary 1, ranking (9) does not imply that country i_1 should only produce and export the high k goods, but instead that it should produce and export relatively more of these goods. This is true irrespective of the number of countries in the economy. Unlike the standard Ricardian model, our model generates a clear and intuitive correspondence between labour productivity and exports. Here, the pattern of comparative advantage for any pair of exporters fully determines their relative export performance across industries.¹⁰

The previous discussion may seem paradoxical. As we have just mentioned, ranking (9) is a weaker version of the ordering at the heart of the standard theory. If so, how does the present theory lead to finer predictions? The answer is simple: it does not. While the standard Ricardian model is concerned with trade flows in each variety of each good, we only are concerned with the total trade flows in each good. Unlike the standard model, we recognize that random shocks—whose origins remain outside the scope of our model—may affect the costs of production of any variety. Yet, by assuming that these shocks are identically distributed across a large number of varieties, we manage to generate clear predictions at the industry level in a multi-country world.¹¹ The lack of such predictions has been the fundamental hindrance to theoretically consistent empirical work on the Ricardian model to date.

Having clarified how the predictions that emerge from our Ricardian model differ from those of the previous theoretical Ricardian literature, we now proceed to tighten the relationship between our model and the real world. That is, we recognize that fundamental productivity levels, which feature in Lemma 1, are not observable.

3.2. Cross-sectional predictions

As in the previous empirical Ricardian literature, our goal is to study the relationship between observed trade flows and observed productivity levels. To do so we assume that statistical agencies perfectly observe $z_i^k(\omega)$ for all varieties of good k produced in country i . What statistical agencies cannot observe, however, is labour productivity for varieties of good k which are not produced in country i because such varieties are being imported from another country.¹² In other words, even in an ideal world without measurement error, statistical agencies cannot report unconditional average productivity, $E[z_i^k(\omega)]$. Instead statistical agencies can only report conditional averages, $E[z_i^k(\omega)|\Omega_i^k]$, based on the set of varieties actually produced in country i , $\Omega_i^k \equiv \cup_{j=1, \dots, I} \Omega_{ij}^k$. From now on, we refer to $\tilde{z}_i^k \equiv E[z_i^k(\omega)|\Omega_i^k]$ as *observed productivity* in country i and industry k and contrast this variable repeatedly with *fundamental productivity*, z_i^k .

Using Lemma 1 and the previous definition, we now describe the impact of observed productivity differences on the cross-sectional variation of aggregate trade flows.

10. Costinot (2009) provides an alternative way to generate clear and intuitive predictions about the pattern of international specialization in a Ricardian model with more than two goods and two countries. The basic idea is to focus on environments in which (i) goods can be ranked in terms of a unique industry characteristic, φ ; (ii) countries can be ranked in terms of a unique characteristic, ψ ; and (iii) productivity is log-supermodular in (φ, ψ) . In such situations, the pattern of international specialization always features a ladder of countries with “high- ψ ” countries specializing in “high- φ ” goods.

11. In this regard, our model bears a resemblance to Davis (1995) in which Ricardian differences also explain intra-industry trade (though factor proportions determine inter-industry trade in his model).

12. This problem was emphasized by Deardorff (1984) in his review of empirical work on the Ricardian model (p. 476): “Problems arise, however, most having to do with the observability of [productivity by industry and country]. The ... problem is implicit in the Ricardian model itself ... [because] the model implies complete specialization in equilibrium ... This in turn means that the differences in labor requirements cannot be observed, since imported goods will almost never be produced in the importing country”.

Theorem 1. Suppose that Assumptions A1–A4 hold. Then for any importer, j , any pair of exporters, i and i' , and any pair of goods, k and k' ,

$$\ln\left(\frac{\tilde{x}_{ij}^k \tilde{x}_{i'j}^{k'}}{\tilde{x}_{ij}^{k'} \tilde{x}_{i'j}^k}\right) = \theta \ln\left(\frac{\tilde{z}_i^k \tilde{z}_{i'}^{k'}}{\tilde{z}_i^{k'} \tilde{z}_{i'}^k}\right) - \theta \ln\left(\frac{d_{ij}^k d_{i'j}^{k'}}{d_{ij}^{k'} d_{i'j}^k}\right), \quad (10)$$

where $\tilde{x}_{ij}^k \equiv x_{ij}^k / \pi_{ii}^k$.

The proof of Theorem 1 proceeds as follows. First, we use Assumption A1 in order to establish the link between observed and fundamental productivity levels:

$$\frac{z_i^k}{z_{i'}^k} = \left(\frac{\tilde{z}_i^k}{\tilde{z}_{i'}^k}\right) \cdot \left(\frac{\pi_{ii}^k}{\pi_{i'i'}^k}\right)^{1/\theta}. \quad (11)$$

The logic behind equation (11) is fairly intuitive. In a given industry k , if country i is more open than country i' , $\pi_{ii}^k \leq \pi_{i'i'}^k$, then country i will tend to produce a smaller, but more productive subset of varieties. Hence, observed relative productivity, $\tilde{z}_i^k / \tilde{z}_{i'}^k$, will be higher than fundamental relative productivity, $z_i^k / z_{i'}^k$. The second term on the right-hand side of equation (11) exactly corrects for this trade-driven selection.¹³ Equation (10) then directly derives from equations (5) and (11).

In our empirical analysis below, we will use Theorem 1 in order to estimate directly the impact (governed by θ) of observed productivity differences on the pattern of trade across countries and industries. Since this empirical exercise is most closely related to previous Ricardian regressions, it is important to highlight how our theory-based approach differs from previous empirical work mentioned in Section 1. It does so in three respects. First, our theory states that the dependent variable in Ricardian regressions should be the log of exports, disaggregated by exporting and importing countries, differenced across exporters and industries, and corrected for differences in levels of “openness” across exporting countries. According to Theorem 1, this new measure of export performance is the theoretically correct way to control for trade-driven selection. Second, our theory clarifies how empirical work can and should control for all the general equilibrium interactions across countries and industries that have a bearing on the partial equilibrium relationship between productivity and exports. Theorem 1 shows that these endogenous relationships do not introduce bias as long as one works with exports and productivity levels that are both in a (log-)“difference-in-difference” form. Third, our micro-theoretical foundations also make it possible to discuss the economic origins of the error term. In our model, these are equal to a particular form of relative trade costs. As a result, our theory makes transparent the plausibility of any orthogonality condition on which the estimation of θ in equation (10) might be based.

3.3. Counterfactual predictions

Theorem 1 offers cross-sectional predictions that will help us to test and quantify the importance of Ricardian comparative advantage in the data. An alternative way of quantifying Ricardian forces is to do counterfactual analysis, *i.e.* to evaluate the effects of moving to a world in which Ricardian forces do not operate across industries. In this section, we pursue such an evaluation

13. Our correction for trade-driven selection in equation (11) is reminiscent of the formula for the welfare gains from trade in Eaton and Kortum (2002). The formal relationship between the two is discussed further in Finicelli, Pagano and Sbracia (2011).

and ask: if, for any pair of exporters, there were no fundamental relative productivity differences across industries, what would be the consequences for aggregate trade flows and welfare?

Formally, we proceed as follows. We start by fixing a reference country i_0 . We assume that fundamental productivity in country i_0 is the same in all sectors in the initial and the counterfactual equilibrium: $(z_{i_0}^k)' \equiv z_{i_0}^k$ for all k . By contrast, for any country $i \neq i_0$ and any sector k , we assume that fundamental productivity in sector k changes from z_i^k to $(z_i^k)'$ such that $(z_i^k)' \equiv Z_i \cdot z_{i_0}^k$, where Z_i is chosen such that the value of the relative wage $(w_i/w_{i_0})'$ in the counterfactual equilibrium is the same as in the initial equilibrium (w_i/w_{i_0}) . We will come back to the exact values of Z_i for all $i \neq i_0$ in a moment. In the rest of this paper, we use the convention $Z_{i_0} \equiv 1$ for the reference country.

It is important to note that there are no fundamental differences in relative productivity across countries in the counterfactual equilibrium. By construction, for any pair of countries, i_1 and i_2 , and any pair of sectors, k_1 and k_2 , we have

$$\frac{(z_{i_1}^{k_1})'}{(z_{i_2}^{k_1})'} = \frac{(z_{i_1}^{k_2})'}{(z_{i_2}^{k_2})'}. \quad (12)$$

In other words, there is no source of Ricardian comparative advantage at the industry level. Furthermore, the adjustment in absolute productivity, Z_i , which is designed to hold relative wages constant around the world, guarantees that changes in fundamental productivity levels from z_i^k to $(z_i^k)'$ have no indirect terms-of-trade effects on the reference country, i_0 . Accordingly, the impact of such changes on aggregate trade flows and welfare in country i_0 can be interpreted as the impact of Ricardian comparative advantage at the industry level.¹⁴ For lack of a better terminology, we will now refer to this particular comparative statics exercise as “Removing country i_0 ’s Ricardian comparative advantage”.

The first step of our counterfactual analysis is to use the trade balance condition, Assumption A5, in order to compute the adjustment in absolute productivity, Z_i , for all $i \neq i_0$.

Lemma 2. *Suppose that Assumptions A1–A5 hold. Adjustments in absolute productivity, $\{Z_i\}_{i \neq i_0}$, can be computed as the solution of the system of equations*

$$\sum_{j=1}^I \sum_{k=1}^K \frac{\pi_{ij}^k (z_i^k/Z_i)^{-\theta} \alpha_j^k \gamma_j}{\sum_{i'=1}^I \pi_{i'j}^k (z_{i'}^k/Z_{i'})^{-\theta}} = \gamma_i, \quad \text{for all } i \neq i_0. \quad (13)$$

An attractive feature of Lemma 2 is that the computation of absolute productivity adjustments, Z_i , requires only a small amount of information. In particular, we do not need data on wages, factor endowments, or trade costs, which may be hard to obtain in practice. In our computation of absolute productivity adjustments (in Section 5.1), we will simply use revealed measures of fundamental productivity z_i^k , together with trade data which enable us to compute export shares, π_{ij}^k , expenditure shares, α_j^k , and world income shares, γ_j .

Let $\hat{X} \equiv X'/X$ denote an arbitrary proportional change in any variable X between the initial and the counterfactual equilibrium. Our last theoretical result can be stated as follows.

Theorem 2. *Suppose that Assumptions A1–A5 hold. If we remove country i_0 ’s Ricardian comparative advantage, then*

14. Another way to see this is to note that starting from an initial equilibrium where equation (12) holds, changes in fundamental productivity levels from z_i^k to $(z_i^k)'$ would have no impact on trade flows and welfare.

1. Counterfactual changes in bilateral trade flows, x_{ij}^k , satisfy

$$\hat{x}_{ij}^k = \frac{(z_i^k/Z_i)^{-\theta}}{\sum_{i'=1}^I \pi_{i'j}^k (z_{i'}^k/Z_{i'})^{-\theta}}, \quad \text{for all } i, j, k. \quad (14)$$

2. Counterfactual changes in country i_0 's welfare, $W_{i_0} \equiv w_{i_0}/p_{i_0}$, satisfy

$$\hat{W}_{i_0} = \prod_{k=1}^K \left[\sum_{i=1}^I \pi_{ii_0}^k \left(\frac{z_i^k}{z_{i_0}^k Z_i} \right)^{-\theta} \right]^{\alpha_{i_0}^k / \theta}. \quad (15)$$

The proof of Theorem 2 is a straightforward generalization of the approach followed by Dekle, Eaton and Kortum (2008) in the one-industry case. Again, only a very small amount of data is needed to compute counterfactual changes in bilateral trade flows and welfare, \hat{x}_{ij}^k and \hat{W}_{i_0} . We still do not need any data on wages, factor endowments, or trade costs. All the economically relevant information about these variables is already included in the π_{ij}^k s.

Equations (14) and (15) are both intuitive. Since our adjustment of absolute productivity levels, Z_i , controls for any indirect terms-of-trade effect, changes in bilateral trade flows, \hat{x}_{ij}^k , should mechanically reflect changes in the productivity of exporter i relative to other exporters i' , $(z_i^k/Z_i)/(z_{i'}^k/Z_{i'})$, weighted by the importance of these exporters in each importing country and industry, $\pi_{i'j}^k$. Similarly, equation (15) states that changes in welfare, \hat{W}_{i_0} , should mechanically reflect the changes in relative prices due to the changes in productivity of different exporters with the π_{ij}^k s providing the economically relevant weights, namely the share of expenditures on goods from country i in country j and industry k .

3.4. Robustness

Our theoretical predictions have been derived under admittedly strong functional form restrictions. In particular, we have assumed, as in Eaton and Kortum (2002), that labour productivity was drawn from an extreme value distribution, Assumption A1. In the working paper version of this paper, Costinot, Donaldson and Komunjer (2010), we have shown, however, that in a world economy featuring small technological differences across countries, our main cross-sectional and counterfactual predictions would hold under weaker restrictions on the distribution of intra-industry productivity shocks. Broadly speaking, the main role of Assumption A1 is to help us go from local to global predictions that hold for arbitrarily large differences across countries and industries.

4. CROSS-SECTIONAL RESULTS

In this section, we investigate how much observed productivity differences affect the cross-sectional variation of bilateral trade flows. That is, we test Theorem 1 using the best available data on internationally comparable productivity and trade flows across countries and industries. This procedure will highlight how empirical tests of the Ricardian model should be performed in a theory-consistent manner.

4.1. Data

Our baseline tests require data of just two types: trade flows and productivity. In this section, we outline the sources of these data.

TABLE 1
Data set description

Sources	OECD STAN Bilateral Trade Database (Edition 2008); GGDC Productivity Level Database: International Comparisons of Output, Inputs and Productivity at the Industry Level (Inklaar and Timmer, 2008).	
Year	1997	
Countries	Australia, Belgium–Luxembourg (aggregated into one country unit to enable a merge with the productivity data), Czech Republic, Denmark, Spain, Finland, France, Germany, Greece, Hungary, Ireland, Italy, Japan, Korea, the Netherlands, Poland, Portugal, Slovakia, Sweden, U.K., and U.S.	
Industry	STAN description	ISIC Rev. 3.1 Code
Food	Food products, beverages, and tobacco	15–16
Textiles	Textiles, textile products, leather, and footwear	17–19
Wood	Wood and products of wood and cork	20
Paper	Pulp, paper, paper products, printing and publishing	21–22
Fuel	Coke, refined petroleum products, and nuclear fuel	23
Chemicals	Chemicals and chemical products	24
Plastic	Rubber and plastics products	25
Minerals	Other non-metallic mineral products	26
Metals	Basic metals and fabricated metal products, except machinery and equipment	27–28
Machinery	Machinery and equipment not elsewhere classified	29
Electrical	Electrical and optical equipment	30–33
Transport	Transport equipment	34–35
Misc. Manuf	Manufacturing not elsewhere classified	36–37

4.1.1. Trade flows. We use trade flow data taken from the OECD Structural Analysis (STAN) Bilateral Trade Database (edition 2008) relating to the year of 1997. Our selection of countries and industries (and the year, 1997) is driven entirely by the availability of both bilateral trade flow data and high-quality productivity data. The resulting sample comprises 21 countries (18 European countries plus Japan, Korea, and the U.S.) and 13 industries which correspond to roughly two-digit ISIC codes and span the manufacturing sector, as described in Table 1.

We use the value of bilateral exports from each of these 21 countries i to each of these 21 countries j in each industry k as our measure of x_{ij}^k .¹⁵ The final source of trade data that we require concerns how much each exporting country i imports in each industry k relative to its total expenditure in that industry, in order to correct for the endogenous selection of varieties that are actually produced domestically, *i.e.* π_{ii}^k in Theorem 1 above. For this purpose, we use data on the import penetration ratio (IPR), which is equal to $1 - \pi_{ii}^k$, of each exporting country i and industry k , as reported in the OECD STAN database.¹⁶

15. Our sample of exporting countries and industries is fundamentally constrained by the intersection of available trade flow and productivity data. However, for many of the exporter–industry pairs in our sample, data on exports are available for many import destinations, not just for imports to another of the 21 countries in our sample. The cross-sectional estimates of θ presented below change only trivially if trade flow data to *all* available import destinations are included in the sample. Our procedure for computing counterfactuals discussed below, however, requires the full (square) matrix of imports and exports, so this procedure can only be performed on the 21 countries in our main sample. It is for this reason that we prefer to use the full square sample throughout.

16. IPR is calculated as imports in the industry divided by the sum of gross output and net imports in the industry. Fifteen exporter–industry observations have an IPR that exceeds one, a fact that presumably reflects re-export activity whose domestic gross output content is recorded in other sectors such as the services sector. We replace these observations with the maximum IPR value, within each exporter, among those values that are less than one.

4.1.2. Productivity. In a Ricardian world, variations in relative productivity levels should be fully reflected in relative producer prices. More specifically, as we formally demonstrate in the online appendix, for any pair of goods, k and k' , and any pair of countries, i and i' , Assumptions A1–A3 imply that

$$\frac{\tilde{z}_i^k \tilde{z}_{i'}^{k'}}{\tilde{z}_{i'}^k \tilde{z}_i^{k'}} = \frac{E[p_{i'}^k(\omega)|\Omega_{i'}^k]E[p_i^{k'}(\omega)|\Omega_i^{k'}]}{E[p_i^k(\omega)|\Omega_i^k]E[p_{i'}^{k'}(\omega)|\Omega_{i'}^{k'}]}. \quad (16)$$

Accordingly, we measure the variation in productivity across countries and industries using differences in producer price indices. Our producer price data are taken from the GGDC Productivity Level Database; see [Inklaar and Timmer \(2008\)](#) for details. In this database, raw price data observations are first collected at the plant level for several thousand products (often with hundreds of products per industry, which we interpret as varieties in the context of our model). This is only made possible due to the use of the PRODCOM system of homogeneous product descriptions within the European Union (EU) and OECD. The GGDC database uses the PRODCOM system to pay particular attention to matching products in different countries in order to control for quality differences. These prices are then aggregated into a unique producer price index at the industry level using output data. We take the inverse of this producer price index as the empirical counterpart of \tilde{z}_i^k in our model.¹⁷

Our relative productivity measures across countries and industries are reported in Table 2. Note that the productivity levels reported in Table 2 are normalized to one in all industries for the U.S. and in all countries for the Food industry. It should be clear that this is without any loss of generality and has no impact on our analysis. At its core, the notion of Ricardian comparative advantage involves the comparison of two goods and two countries, which is reflected in the fact that predicted export behaviour (in Theorem 1) depends only on the difference-in-difference in $\ln \tilde{z}_i^k$.

4.2. Baseline results

Following Theorem 1, we estimate the following log-linear model using the data described above:

$$\ln\left(\frac{\tilde{x}_{ij}^k \tilde{x}_{i'j}^{k'}}{\tilde{x}_{i'j}^{k'} \tilde{x}_{ij}^k}\right) = \theta \ln\left(\frac{\tilde{z}_i^k \tilde{z}_{i'}^{k'}}{\tilde{z}_{i'}^k \tilde{z}_i^{k'}}\right) + \ln\left(\frac{\varepsilon_{ij}^k \varepsilon_{i'j}^{k'}}{\varepsilon_{i'j}^{k'} \varepsilon_{ij}^k}\right), \quad (17)$$

where $\ln \tilde{x}_{ij}^k \equiv \ln x_{ij}^k - \ln \pi_{ii}^k$ is (log) “corrected exports” from country i to country j in industry k , or (log) exports adjusted for openness in country i and industry k , $\ln \pi_{ii}^k$; and $\ln \tilde{z}_i^k$ is (log) observed productivity. The error term ε_{ij}^k captures both the variation in (log) variable trade costs, $-\theta \ln d_{ij}^k$, as well as any measurement error in bilateral trade flows.

As mentioned earlier, the (log) difference-in-difference in equation (10) and its empirical counterpart, equation (17) above, highlight the essence of trade based on Ricardian comparative advantage. That is, relative productivity differences across any pair of countries and industries drive relative export levels to any market j . While it is possible to estimate equation (17) in

17. It is worth pointing out that producer price indices from the GGDC database relate to gross output. We believe that such measures are well suited for our empirical exercise for two reasons. First, as we discuss below, our theory is fundamentally one of comparative costs. When perfectly competitive producers (like those in our model) use intermediate inputs in production, it is the prices of their gross output that are equal to their unit costs. Second, the GGDC database is primarily concerned with collecting internationally comparable output rather than input prices, so we expect producer prices of gross output to be the best measured proxy for productivity available. Nevertheless, for robustness, Section 4.3 below considers an extension that aims to control for variation in input prices (and input intensity).

TABLE 2
Relative productivity levels, by country and industry

	Food	Textiles	Wood	Paper	Fuel	Chemicals	Plastic	Minerals	Metals	Machinery	Electrical	Transport	Misc.Manuf
AUS	1	0.77	0.89	0.85	1.21	0.84	0.81	1.00	0.84	0.86	0.82	0.74	0.84
BEL	1	0.84	0.84	1.03	1.70	1.26	1.71	1.49	1.07	0.93	0.86	0.71	1.11
CZE	1	0.80	1.16	0.72	0.66	1.44	1.65	1.88	1.05	0.97	0.93	0.78	0.85
DEN	1	0.75	0.84	0.63	0.64	1.02	1.01	0.98	0.85	0.85	0.63	0.51	1.35
ESP	1	0.92	1.47	0.91	0.70	1.34	1.38	1.58	1.08	0.99	1.08	1.01	1.15
FIN	1	0.82	1.20	1.18	0.86	1.69	1.37	1.30	1.18	1.13	0.90	0.65	1.47
FRA	1	0.76	1.22	0.93	1.11	1.24	1.64	1.28	1.08	1.13	0.93	0.84	0.92
GER	1	0.67	0.91	0.89	0.94	1.03	1.23	1.20	0.91	0.92	0.80	0.72	0.90
GRE	1	1.07	1.65	1.57	1.13	1.87	1.21	2.19	1.51	1.54	1.26	0.83	1.36
HUN	1	0.80	1.05	1.08	0.52	0.98	1.32	1.33	0.91	1.52	0.83	0.49	1.14
IRE	1	0.93	1.03	0.81	1.15	1.49	1.55	1.19	1.11	0.92	0.74	0.62	1.11
ITA	1	0.79	1.02	0.68	0.65	1.26	1.48	1.63	1.17	1.02	0.98	1.14	0.94
JPN	1	1.88	1.37	1.69	1.06	1.84	1.69	1.71	1.85	2.05	1.96	1.96	1.33
KOR	1	1.23	1.56	1.40	1.32	2.74	1.97	2.22	1.50	1.64	1.60	1.72	1.94
NED	1	0.85	0.54	0.76	0.65	1.25	1.26	1.11	1.00	0.71	0.75	0.61	0.74
POL	1	1.15	1.12	0.92	1.07	1.24	2.21	1.59	0.91	0.80	0.60	0.68	0.64
PTL	1	1.18	1.32	1.08	0.62	1.26	2.32	1.66	1.42	1.01	1.01	0.59	0.91
SLK	1	0.63	0.82	0.91	0.49	1.04	1.07	1.45	0.71	0.63	0.72	0.65	0.89
SWE	1	0.68	1.18	0.92	0.78	1.22	1.18	0.92	1.14	0.98	0.89	0.67	1.09
U.K.	1	0.92	0.69	1.23	1.27	1.20	1.74	1.24	1.02	1.29	0.97	0.79	1.80
U.S.	1	1	1	1	1	1	1	1	1	1	1	1	1

Notes: Country and industry productivity, as measured by the inverse of producer prices (as collected by the GGDG Productivity Level Database). Entries are normalized to reflect relative productivity levels in the manner suggested by equation (10). That is, within each industry the U.S. has productivity equal to 1, and within each country, the "Food" industry (ISIC Rev 3.1 code: 15–16) has productivity equal to 1. Countries are listed in the order presented in Table 1.

TABLE 3
Cross-sectional results—baseline

Dependent variable	log (corrected exports)	log (exports)	log (corrected exports)	log (exports)
	(1)	(2)	(3)	(4)
log (productivity based on producer prices)	1.123*** (0.0994)	1.361*** (0.103)	6.534*** (0.708)	11.10*** (0.981)
Estimation method	OLS	OLS	IV	IV
Exporter \times importer fixed effects	YES	YES	YES	YES
Industry \times importer fixed effects	YES	YES	YES	YES
Observations	5652	5652	5576	5576
R^2	0.856	0.844	0.747	0.460

Notes: Regressions estimating equation (18) using data from 21 countries and 13 manufacturing sectors (listed in Table 1) in 1997. “Exports” is the value of bilateral exports from the exporting country to the importing country in a given industry. “Corrected exports” is “exports” divided by the share of the exporting country’s total expenditure in the given industry that is sourced domestically (equal to one minus the country and industry’s IPR). “Productivity based on producer prices” is the inverse of the average producer price in an exporter–industry. Columns (3) and (4) use the log of 1997 R&D expenditure as an instrument for productivity. Data sources and construction are described in full in Section 4.1. Heteroskedasticity-robust standard errors are reported in parentheses. ***Statistically significantly different from zero at the 1% level.

this form, it is simpler—and closer to existing empirical work—if we estimate the following econometrically equivalent specification:

$$\ln \tilde{x}_{ij}^k = \delta_{ij} + \delta_j^k + \theta \ln \tilde{z}_i^k + \varepsilon_{ij}^k. \quad (18)$$

In this expression, δ_{ij} represents an importer–exporter fixed effect and δ_j^k an importer–industry fixed effect.¹⁸ Under the assumption that variable trade costs (and other components of the error term, ε_{ij}^k) are orthogonal to observed productivity, an OLS estimate of equation (18) provides an unbiased estimate of θ , the extent of intra-industry heterogeneity in this model. We come back to the plausibility of this orthogonality restriction in a moment.

The first column of Table 3 reports a preliminary estimate of θ , from estimating equation (18) by OLS. In line with the prediction in Theorem 1, this estimate is positive and statistically significant.¹⁹ According to this estimate of the productivity-to-exports elasticity, a 1% change in productivity is, all else held equal, associated with a 1.1% change in exports.

Column (2) of Table 3 reports the OLS estimate of θ if the dependent variable is not adjusted for the difference between fundamental and observed productivity highlighted by Theorem 1. Without this adjustment, we see that one would tend to overestimate the importance of productivity differences. This is intuitive. Observed productivity differences are smaller than fundamental productivity differences since countries with low fundamental productivity levels only produce varieties for which they get very good productivity draws. Thus without our

18. Strictly speaking, equations (17) and (18) only are econometrically equivalent for balanced panels. Since there are missing observations in our data set, the fixed-effect estimator therefore provides an average of all possible difference-in-difference estimators (whose actual values depend on the reference country and industry). Note also that since we have a full set of importer–exporter fixed effects, δ_{ij} , our estimation strategy subsumes exporter fixed effects and importer fixed effects.

19. The standard errors reported in Tables 3 and 4 are adjusted for unrestricted forms of heteroskedasticity. Standard errors that are clustered at the exporter–industry level are larger as one would expect if the error term ε_{ij}^k in equation (18) is correlated across exporter destinations. But our preferred estimate of θ is still statistically significant at standard levels when clustering at the exporter–industry level.

correction, one would wrongly infer that smaller productivity differences lead to the same variation in trade flows, thereby overestimating the elasticity θ of trade flows with respect to productivity.

There are two main reasons to be concerned with the previous OLS estimates of equation (18) introduced above: (i) simultaneity bias, due to agglomeration effects through which higher export levels lead to higher productivity levels and (ii) attenuation bias, due to measurement error in productivity. To attempt to circumvent these two potential sources of bias, we estimate equation (18) by the method of instrumental variables, with the endogenous regressor—productivity levels ($\ln \tilde{z}_i^k$)—instrumented with (the log of) research and development (R&D) expenditures at the country–industry level.²⁰ This approach follows Eaton and Kortum (2002) and Griffith, Redding and Van Reenen (2004) in modelling technology as a function of R&D activity. In doing so, our identifying assumption is that relative R&D expenditures are correlated with relative bilateral trade flows only through their impact on relative productivity, *i.e.* relative producer prices.

Our IV estimate of θ is reported in Column (3) of Table 3.²¹ This is our preferred specification. Compared to the OLS estimates, the magnitude of θ is considerably larger—6.53 rather than 1.12—and still statistically significant. This increase is consistent with the likely scenario in which our OLS estimates suffer from attenuation bias because producer prices are extremely difficult to measure accurately in practice.²² Finally, in Column (4) of Table 3 we see that if we were not adjusting for the distinction between fundamental and observed productivity, the value of our estimate of θ would be significantly biased upwards (from 6.53 to over 11) in both statistical and economic senses.²³

Taken together, our findings in Table 3 highlight the importance of the endogeneity of relative productivity (and hence the importance of IV estimation relative to OLS estimation) and the importance of correcting for the trade-driven wedge between observed and fundamental productivity levels, as emphasized in Theorem 1.

In Sections 4.3 and 4.4 below, we document that our core finding (in Column (3) of Table 3) is robust to a variety of specification checks. Before moving on to these robustness checks, however, it is important to note that our preferred estimate of $\theta = 6.53$ is in line with previous estimates of θ obtained by researchers using different methodologies. In particular, Eaton and Kortum (2002) estimate θ to be 3.60 using wage data and either 8.28 or 12.86 using data on price gaps between countries (to proxy for trade costs); Simonovska and Waugh (2009) amend the price gap methodology of Eaton and Kortum (2002) (using both the price data of Eaton and Kortum (2002) and an improved round of the underlying ICP price data) and estimate θ to be

20. Data on R&D expenditure are from the Analytical Business Enterprise Research and Development (ANBERD) database collected by the OECD—as used, *e.g.* by Griffith, Redding and Van Reenen (2004). Wherever possible this source aims to break down the R&D expenditures of large, multi-industry firms, as well as those of enterprise-serving research institutes, into expenditures by output product. Sixteen exporter–industry observations have missing R&D expenditure information in this data source. We interpolate these observations using the fitted values from a regression on exporter and industry fixed effects.

21. The instrument has strong predictive power. When run only across exporters and industries (*i.e.* utilizing only the meaningful variation in the first stage), the heteroskedasticity-robust *t*-statistic in the first stage is equal to 2.23.

22. As discussed in Costinot, Donaldson and Komunjer (2010), agglomeration effects have *a priori* ambiguous effects on our estimates of θ .

23. By statistical sense, we formally mean the following. If we estimate a full three-equation system (in which equation (1) is the first-stage relationship between observed productivity and R&D expenditure, equation (2) is the second-stage relationship estimated in Column (3) of Table 3, and equation (3) is the second-stage relationship estimated in Column (4) of Table 3) that allows for correlation among the error terms across equations, the chi-squared test for equality of the coefficient estimate of θ in equation 2 and the estimate of θ in equation 3 has a *p*-value that is smaller than 0.001.

4.5 regardless of the sample. Finally, Donaldson (2008) estimates θ commodity-by-commodity in colonial India using a trade costs approach and finds an average (over 85 commodities) of 5.2, with a standard deviation of 2.1.

4.3. Alternative productivity measures

In the previous subsection, we have taken the Ricardian model developed in Section 2 very seriously, interpreting differences in producer prices as differences in productivity. There are good reasons to believe that the variation in producer prices may reflect more than productivity differences in practice. In particular, variation in producer prices may also capture differences in factor prices and factor intensity across countries and industries, as emphasized in the Heckscher–Ohlin model. The objective of this subsection is to discuss how the introduction of additional factors of production would affect our empirical analysis and the interpretation of our empirical results.

The first thing to note is that our main specification, equation (18), does not rely on the assumption that there only is one factor of production (or that there are no intermediate inputs to production). Our theory is fundamentally one of comparative costs. These costs relate one-for-one with producer prices in a perfectly competitive setting, regardless of whether the determinants of production costs are productivity or factor price differences. With multiple factors of production (or intermediate inputs), the relationship between bilateral imports and producer prices would still be given by equation (18), with θ now representing intra-industry heterogeneity in terms of total factor productivity (TFP). Thus, the only issue related to the previous empirical results is one of interpretation, to which we now turn.

Econometrically, our IV estimates reflect the impact of predicted changes in producer prices conditional on R&D expenditures. If R&D expenditures affect producer prices only through changes in TFP, our IV estimates should therefore capture, in line with our Ricardian interpretation, the impact of TFP differences on bilateral trade flows. In order to assess the validity of this interpretation formally, we estimate θ using a measure of TFP rather than producer price data. Since the GGDC Productivity Level Database also reports data on the prices and shares of labour, capital, and material inputs, (log-)TFP measures can be computed in the standard “dual” fashion as the (log-)difference between the producer price and the sum of (log-)factor prices weighted by their factor shares.²⁴

Having constructed this dual TFP measure of productivity in order to control for factor price and intensity differences, we then regress exports on productivity, as in equation (18), using this TFP measure of productivity rather than producer prices as before. The new IV estimate of θ in this regression is reported in Column (2) of Table 4. While the estimate of θ increases slightly to 6.70 (when compared with our previous estimate of 6.53, reported again in Column (1) of Table 4 for convenience), it is still positive and significantly different from zero, in line with Theorem 1. Further, this new estimate lies comfortably within the 95% confidence interval of our previous estimate (and also within the bounds of estimates of θ previously obtained in the literature, as discussed above). Both of these findings lend strong support to our Ricardian interpretation of the results presented in Table 3. Although our use of relative producer prices to measure relative productivity is *a priori* consistent with any theory of comparative costs—regardless of the number of non-tradable factors of production—the estimates in Columns (1) and (2) of Table 4 imply that Ricardian motives for cross-industry trade (at least those contained in the

24. That is, our “dual” (log-)TFP measure is calculated as $\ln TFP_i^k = \ln p_i^k - \alpha_{i,L}^k \ln w_i^k - \alpha_{i,K}^k \ln r_i^k - \alpha_{i,M}^k \ln m_i^k$, where the variables α refer to factor shares of labour, capital, and material inputs, respectively, and w_i^k , r_i^k , and m_i^k refer to the wage rate, capital rental rate, and material inputs price, respectively.

TABLE 4
Cross-sectional results—alternative productivity measures

Dependent variable in all regressions: corrected exports				
	(1)	(2)	(3)	(4)
log(productivity based on producer prices)	6.534*** (0.708)			
log(productivity based on producer prices, dual TFP measure)		6.704*** (0.874)		
log(productivity based on real gross output per worker)			2.725*** (0.234)	
log(productivity based on real gross output, primal TFP measure)				4.316*** (0.504)
Observations	5576	5576	5576	4541
R^2	0.747	0.587	0.839	0.835

Notes: Regressions are based on estimating equation (18) using data from 21 countries and 13 manufacturing sectors (listed in Table 1) in 1997. All regressions estimated use the log of 1997 R&D expenditure as an instrument for the productivity measure and with exporter-times-importer and industry-times-importer fixed effects included. “Corrected exports” is the value of bilateral exports divided by the share of the exporting country’s total expenditure in the given industry that is sourced domestically (equal to one minus the country and industry’s IPR). “Productivity based on producer prices” is the inverse of the average producer price in an exporter–industry. “Productivity based on producer prices, dual TFP measure” is the inverse of the average producer price divided by an income share-weighted composite of producer input prices (wages, capital rental rates, and intermediate inputs), as defined in footnote 24. “Productivity based on real gross output per worker” is a nominal measure of gross output, deflated by average gross output producer price indices. “Productivity based on real gross output, primal TFP measure” is the nominal value of gross output deflated by a producer gross output price index, divided by an income share-weighted composite of input values deflated by input price indices, as defined in footnote 26. Data sources and construction are described in full in Section 4.1. Heteroskedasticity-robust standard errors are reported in parentheses. ***Statistically significantly different from zero at the 1% level.

component of productivity that is correlated with R&D expenditure) appear to be orthogonal to Heckscher–Ohlin motives in our data.²⁵

The remaining two columns of Table 4 report estimates from regressions in which two alternative proxies for productivity are used. These alternatives are all variations of “primal” measures of labour and multi-factor productivity: real gross output per worker and TFP calculated in the primal manner rather than the dual manner introduced above.²⁶ Each column contains the resulting estimate of θ from a separate regression, each of which is an IV regression of equation (18) using one of the above alternative productivity proxies as a measure of \tilde{z}_i^k .

Estimates of θ based on these alternative productivity measures are both somewhat lower than our preferred estimate (in Column (1)). This is consistent with the idea that these alternative productivity proxies are mismeasured (and that this measurement error problem is not entirely obviated by our instrumental variable procedure). As we have argued above, in a Ricardian environment relative producer prices contain all the relevant information about relative

25. This finding echoes the work of Morrow (2010).

26. These alternative productivity proxies are reported in the GGDC database and are constructed in a standard fashion. That is, real gross output per worker is constructed as the nominal value of gross output per worker in an exporter–industry divided by the producer price index and “primal” TFP is calculated as $\ln \text{TFP}_i^k = \ln Y_i^k - \alpha_{iK}^k \ln K_i^k - \alpha_{iL}^k \ln L_i^k - \alpha_{iM}^k \ln M_i^k$, where Y_i^k , K_i^k , L_i^k , and M_i^k represent real values of gross output, capital, labour, and intermediate inputs, respectively (where these real values are computed as the ratio of nominal values over the relevant price index), and the α terms represent income shares.

productivity levels. Because our two alternative productivity measures are constructed using data on producer prices plus additional variables (such as nominal output and the number of workers in the case of real output per worker), these two measures are effectively our preferred productivity measure plus some variables carrying additional scope for measurement error.²⁷ That is, in the light of a Ricardian model, it is natural that Column (1) contains a larger estimate of θ than Columns (3) and (4) because it is based on the measure of productivity that is least likely to be mismeasured.

4.4. *Alternative samples*

In this section, we explore the extent to which our estimate of $\theta = 6.53$, obtained above, is robust to a number of alternative samples. The first sample restriction we consider is to country pairs in the EU only, which we use to test how our results depend on our assumption of orthogonality between relative trade costs and relative productivity. The second set of sample restrictions we consider is to observations for which we have increased confidence that producer prices were collected in a manner consistent with our theory. In all these sample restrictions, our estimate of θ is reassuringly robust (varying from 4.62 to 8.06) and within the bounds of previous estimates in the literature.

4.4.1. Endogenous trade protection. In Section 4.2, we have described two potential sources of bias of our OLS estimates and discussed how they could be addressed by an IV strategy. This IV strategy, however, does not obviate issues related to endogenous trade protection, which could bias our estimates downward (see Costinot, Donaldson and Komunjer, 2010). In order to address this issue, we now estimate equation (18) again by IV but on a sample that includes EU exporters and importers only. Within this free trade area, there are no trade barriers and, hence, no scope for endogenous trade protection.

The results from restricting our sample to EU members are reported in Column (2) of Table 5. Our preferred estimate of θ is reasonably stable to this sample restriction: it falls to 4.62, which still lies within the 95 % confidence interval of our previous estimate of 6.53 (reported again in Column (1) of Table 5 for convenience) and remains statistically significant despite the reduction in sample size. It therefore appears that our estimates are unlikely to be severely biased due to endogenous trade protection.

4.4.2. Producer vs. expenditure prices. In an open economy, goods price data are only informative about domestic productivity differences to the extent that they reflect domestic producer rather than expenditure prices. Unfortunately, though most of the GGDC's internationally comparable price estimates are based on aggregations of micro-level observations of producer prices, in order to achieve a complete sample the GGDC were forced to include some observations of expenditure prices. Helpfully, however, the GGDC has published data on the share of raw price observations, within each country and industry, that were obtained from producer rather than expenditure accounts.²⁸ We can therefore use this information to explore how our

27. Consider the example of real gross output per worker, a productivity measure that has been used extensively in the empirical Ricardian literature. This variable is computed as nominal gross output per worker divided by a producer price index (*i.e.* by our preferred productivity measure). Since nominal gross output per worker should be equal—in our Ricardian setting—to the nominal wage which is equal across sectors, any divergence between relative real gross output per worker and relative (inverse) producer prices per worker must be due to measurement error in nominal output per worker (either because of measurement error in nominal output or the number of workers).

28. This information is tabulated in Appendix Table 3 of Inklaar and Timmer (2008). We have digitized the content of these tables so that it is included in the data set used in the present paper.

TABLE 5
Cross-sectional results—alternative samples

	Dependent variable in all regressions: corrected exports			
	Entire sample	EU member countries only	Producer price data below median quality	Producer price data above median quality
	(1)	(2)	(3)	(4)
log (productivity based on producer prices)	6.534*** (0.708)	4.621*** (0.585)	5.820*** (1.106)	8.057*** (1.267)
Observations	5576	2162	2795	2781
R^2	0.747	0.808	0.820	0.688

Notes: Regressions are based on estimating equation (18) using data from 21 countries and 13 manufacturing sectors (listed in Table 1) in 1997. All regressions instrument for productivity using log R&D expenditures in 1997 and include exporter-times-importer fixed effects and industry-times-importer fixed effects. “Corrected exports” is the value of bilateral exports divided by the share of the exporting country’s absorption in the given industry that is sourced domestically (equal to one minus the country and industry’s IPR). “Productivity based on producer prices” is the inverse of the average producer price in an exporter–industry. “Producer price data quality” is based on the percentage of underlying price data observations within the exporter–industry that were obtained (in the creation of the GGDC data set) using producer price data rather than expenditure price data. Data sources and construction are described in full in Section 4.1. Heteroskedasticity-robust standard errors are reported in parentheses. ***Statistically significantly different from zero at the 1% level.

estimates change as we restrict our sample to the observations that are best suited to measuring producer productivity—namely, those that are drawn from underlying observations of producer rather than expenditure prices. In particular, we can rank the exporter–industry observations by the percentage of price data based on producer prices and split our sample into groups of “price data quality” based on this ranking.

Our previous results (in Tables 3 and 4) were based on the entire sample of price data.²⁹ By contrast, Column (3) of Table 5 reports IV results based on two different samples that are partitioned based on the quality of their producer price data—Column (3) reports results for the sample that contains only exporter–industry observations with below-median price data quality and Column (4) presents those with above-median price data quality. Our preferred estimate of θ is again reassuringly stable—given the standard errors on these estimates—to this sample truncation, changing from 6.53 (whole sample in Column (1)) to 5.82 (below-median quality sample in Column (3)) to 8.06 (above-median quality sample in Column (4)). Overall, we believe that the cluster of estimates of θ presented in Tables 4 and 5 adds considerable confidence to our preferred estimate of $\theta = 6.53$.

4.5. Summary

What have we learned from the cross-sectional results in Tables 3–5? Our findings can be summarized as follows:

1. The theoretical prediction of Theorem 1 is consistent with the data: relative export levels across countries and industries, corrected for trade-driven selection, are positively correlated with relative productivity levels across countries and industries.

29. The median observation in the unrestricted sample, used in Column (1) of Table 5, is based on underlying price observations that are composed of 84 % producer prices and 16 % expenditure prices.

2. Correcting for trade-driven selection is quantitatively important: failure to correct export flows in this manner leads to bias that is both substantial, from 6.53 to 11.1, and in the direction suggested by our Ricardian model.
3. Our preferred estimate of $\theta = 6.53$ —which captures intra-industry heterogeneity à la Eaton and Kortum (2002)—is comfortably in line with previous estimates of this coefficient obtained in the literature using different methodologies.
4. The difference between our OLS and IV estimates of θ suggests measurement error in international productivity levels, which resonates well with the numerous challenges of measuring productivity consistently across countries and industries.
5. Our estimate of θ is extremely similar, $\theta = 6.70$, when using a productivity measure that strips out variation in factor intensity and factor prices across countries and industries, thereby giving support to the Ricardian interpretation of our results.
6. Our estimates of θ are robust to a number of sample restrictions that alleviate, among other things, concerns of bias due to endogenous trade protection.

Taken together these empirical results provide support for the ability of the Ricardian model to explain the pattern of trade across countries and industries. Comforted by such results, we go on to ask: In today's world trading system, how important—in a well-defined, welfare-based sense—are Ricardian forces?

5. COUNTERFACTUAL RESULTS

In order to investigate further the quantitative importance of productivity differences, we use Theorem 2 to ask the following counterfactual questions: What if, for any pair of exporters, there were no fundamental relative productivity differences across industries? What would be the consequences for aggregate trade flows and welfare? As mentioned earlier, we refer to this counterfactual exercise as “Removing a country's Ricardian comparative advantage”, though it is important to stress that in our model there are also Ricardian motives *within* each industry that our counterfactual exercise will leave intact.

5.1. Preliminary: revealed productivity

The formal procedure to compute our counterfactuals is as described in Section 3.3. However, to compute the initial productivity levels, z_i^k , that enter equation (13), we do not use producer price data from the GGDC Productivity Level Database, which, as argued in Section 4, suffer from severe measurement error.³⁰ Instead, we use our model to obtain “revealed” measures of productivity at the country and industry level. Formally, we estimate

$$\ln x_{ij}^k = \delta_{ij} + \delta_j^k + \delta_i^k + \varepsilon_{ij}^k, \quad (19)$$

where δ_{ij} , δ_j^k , and δ_i^k are exporter–importer, importer–industry, and exporter–industry fixed effects, respectively. According to our model, bilateral trade flows satisfy

$$\ln x_{ij}^k = \delta_{ij} + \delta_j^k + \theta \ln z_i^k + \varepsilon_{ij}^k.$$

Hence, estimates of δ_i^k can be used to construct revealed measures of productivity, $e^{\delta_i^k/\theta}$, in country i and industry k . These measures, which will be used in all of our counterfactual procedures

30. Measurement error did not affect our IV estimates of θ . Our IV strategy was precisely designed to overcome measurement error and obtain coefficient estimates that are purged of attenuation bias.

below, are reported in Table 6.³¹ As in Table 2, productivity levels are normalized to one in all industries for the U.S. and in all countries for the Food industry.³²

Before turning to our counterfactual results, it is worth pointing out that this simple way of computing revealed measures of productivity, while not the main focus of our paper, provides a theoretically consistent alternative to Balassa's (1965) well-known index of "revealed comparative advantage". In Balassa (1965), the revealed comparative advantage of country i in industry k is formally defined as

$$\left(x_{i\text{World}}^k / \sum_{k'=1}^K x_{i\text{World}}^{k'} \right) / \left(\sum_{i'=1}^I x_{i'\text{World}}^k / \sum_{i'=1}^I \sum_{k'=1}^K x_{i'\text{World}}^{k'} \right),$$

where $x_{i\text{World}}^k$ are the total exports of country i in industry k . Like Balassa (1965), we therefore offer a methodology that uses data on relative exports to infer the underlying pattern of comparative advantage across countries and industries. There are, however, two important differences between Balassa's (1965) approach and ours. First, our empirical strategy is theoretically grounded. The ranking of the OLS estimates of $(\delta_{\text{US}}^k - \delta_i^k)$, for instance, is the empirical counterpart to the ranking of $(\ln z_i^k - \ln z_{\text{US}}^k)$ in our model. By contrast, Balassa's index would not, in general, provide a ranking of relative productivity levels in our model. Second, our approach fundamentally is about the pairwise comparisons across exporters and industries that are at the core of comparative advantage in a Ricardian world. Put simply, our fixed effects aim to uncover which of Portugal and England is the country relatively better at producing wine than cloth. They do not try to assess whether Portugal is good at producing wine compared to an intuitive but *ad hoc* benchmark. Unlike Balassa (1965), we do not aggregate exports across countries and industries, which allows us to separate the impact of productivity differences from trade costs and demand differences.

5.2. Trade flows

We first consider the impact of "removing a country's Ricardian comparative advantage" on trade flows. For each country and industry, we compute changes in total exports (to all destination countries combined) using equation (14). The first column of Table 7 describes the change in total exports for the reference country, *i.e.* the country whose productivity remains unchanged. The overall picture is clear: on average, changes in total trade volumes are small. This should not come as a surprise. By construction, wages, and hence GDP levels, are unchanged in the counterfactual equilibrium. Thus, total trade volumes, which are related to the dispersion of GDP levels across countries, do not change very much either on average.

Intuitively, the main impact of our counterfactual exercise should be on the share of inter-industry trade. If there were no trade costs and no differences in preferences across countries, then "removing a country's Ricardian comparative advantage" would eliminate all inter-industry trade, *i.e.* trade would be balanced industry by industry. To explore this idea systematically, we use a linear transformation of the Grubel and Lloyd (1975) index. For each country i and industry

31. Using revealed productivity measures does not alter our main results dramatically. If we use observed productivity (measured as inverse producer prices) projected on our instrumental variable (R&D expenditure), our most important result, the world average welfare loss from moving "removing a country's Ricardian comparative advantage", is 2.3% rather than 5.3% of the total gains from trade. As in Section 5.3, there also is considerable heterogeneity in this number across countries.

32. Of course, our fixed-effect regression only identifies $(\delta_i^k - \delta_i^{k'}) - (\delta_{i'}^k - \delta_{i'}^{k'})$, and therefore, $(\ln z_i^k - \ln z_i^{k'}) - (\ln z_{i'}^k - \ln z_{i'}^{k'})$. These differences-in-differences, however, capture the essence of Ricardian comparative advantage and are all that is necessary for our counterfactual predictions.

TABLE 6
Revealed relative productivity levels, by country and industry

	Food	Textiles	Wood	Paper	Fuel	Chemicals	Plastic	Minerals	Metals	Machinery	Electrical	Transport	Misc. Manuf
AUS	1	0.91	0.73	0.63	0.74	0.75	0.74	0.69	0.88	0.71	0.71	0.68	0.77
BEL	1	1.16	0.87	0.91	1.03	1.06	1.07	1.07	1.12	0.89	0.80	0.89	0.98
CZE	1	1.37	1.16	1.07	0.93	1.05	1.21	1.56	1.41	1.13	0.90	1.10	1.23
DEN	1	0.90	0.85	0.77	0.65	0.77	0.90	0.82	0.89	0.83	0.72	0.63	0.86
ESP	1	1.14	0.86	0.86	0.81	0.87	1.03	1.16	1.08	0.83	0.76	0.96	0.94
FIN	1	1.09	1.54	1.70	0.97	1.04	1.17	1.19	1.46	1.17	1.19	0.90	1.09
FRA	1	1.08	0.83	0.91	0.97	0.97	1.00	1.06	1.07	0.88	0.85	0.90	0.90
GER	1	1.20	0.96	1.05	0.98	1.06	1.20	1.17	1.26	1.09	0.97	1.08	1.04
GRE	1	1.16	0.59	0.74	0.86	0.72	0.83	1.00	0.97	0.65	0.59	0.50	0.75
HUN	1	1.13	0.89	0.84	0.79	0.89	0.98	1.06	1.06	0.82	0.87	0.84	0.82
IRE	1	0.83	0.64	1.08	0.57	0.98	0.85	0.76	0.82	0.72	0.92	0.55	0.82
ITA	1	1.44	0.92	0.96	0.82	0.97	1.16	1.29	1.21	1.10	0.86	0.92	1.22
JPN	1	1.51	0.87	1.31	1.10	1.67	1.84	1.75	1.75	1.80	1.81	1.93	1.72
KOR	1	1.69	0.84	0.92	0.87	1.20	1.46	1.04	1.36	1.21	1.32	1.42	1.34
NED	1	0.96	0.68	0.87	0.95	0.89	0.92	0.87	0.96	0.77	0.81	0.74	0.84
POL	1	1.18	1.06	0.91	0.99	0.92	0.99	1.18	1.26	0.86	0.83	0.84	1.06
PTL	1	1.55	1.32	1.03	0.79	0.87	1.05	1.31	1.05	0.89	0.90	0.95	0.94
SLK	1	1.33	1.24	1.14	0.77	1.13	1.32	1.48	1.63	1.09	0.93	0.91	1.16
SWE	1	1.12	1.37	1.39	1.16	1.08	1.24	1.15	1.40	1.14	1.09	1.04	1.12
U.K.	1	1.16	0.75	1.03	1.09	1.01	1.08	1.09	1.14	0.96	0.95	0.93	0.99
U.S.	1	1	1	1	1	1	1	1	1	1	1	1	1

Notes: Country and industry productivity, as measured by estimates of exporter–industry fixed effects in the estimation of equation (19). Entries are normalized to reflect relative productivity levels in the manner suggested by equation (10). That is, within each industry the U.S. has productivity equal to 1, and within each country, the “Food” industry (ISIC Rev 3.1 code: 15–16) has productivity equal to 1. Countries are listed in the order presented in Table 1.

TABLE 7
Counterfactual results—baseline

Reference country	Outcome variable of interest			
	% change in in total exports	Change in index of interindustry trade	% change in welfare	% change in welfare relative to the total gains from trade
	(1)	(2)	(3)	(4)
Australia	18.52	24.57	-2.90	-39.11
Belgium and Luxembourg	-1.76	4.12	0.71	2.64
Czech Republic	3.91	5.62	-0.12	-1.26
Denmark	0.60	-2.64	-0.40	-2.18
Spain	3.68	-3.89	-0.46	-7.08
Finland	-5.62	3.44	0.14	1.65
France	0.80	-0.49	-0.20	-3.09
Germany	-2.10	-8.46	0.14	2.22
Greece	26.35	-11.23	-4.37	-40.47
Hungary	1.70	-5.28	-0.25	-1.62
Ireland	-5.48	-4.31	0.20	0.74
Italy	-4.76	-9.85	0.14	2.78
Japan	-6.12	-24.75	0.35	24.48
Korea	2.68	-10.15	-0.44	-9.60
Netherlands	1.95	-0.94	-0.64	-2.81
Poland	12.33	-22.35	-1.68	-23.09
Portugal	8.44	-13.62	-0.92	-9.12
Slovakia	2.33	14.11	0.82	4.64
Sweden	-2.98	3.03	0.34	3.30
U.K.	3.45	-4.04	-0.26	-2.94
U.S.	3.82	-3.83	-0.42	-11.71
World average	2.94	-5.72	-0.49	-5.32

Notes: Results from counterfactual calculations in which, one at a time for each country listed (the “reference country”), every other country in the world is given the reference country’s relative productivity levels across industries, while adjusting each country’s absolute productivity levels in such a way as to hold nominal wages fixed around the world (so as to neutralize terms-of-trade effects). The methodology follows Lemma 4 and Theorem 5. Column (1) reports the resulting proportional change in the total volume of the reference country’s exports. Column (2) reports the resulting change in the reference country’s index of inter-industry trade (given in Section 5.2); this index takes the value of 100 in the case of pure inter-industry trade and zero in the case of pure intra-industry trade. Column (3) reports the resulting proportional change in welfare (equal to real income spent on manufacturing) for the reference country; a negative number indicates a welfare loss. And Column (4) reports this change in welfare as a percentage of the total gains from trade; a negative number indicates a welfare loss (and a value of 100 means that this loss is equal to that of moving to autarky). The row labelled “World average” reports the unweighted average of the country-specific results above.

k , we compute $100 \times \sum_{j \neq i} |x_{ij}^k - x_{ji}^k| / \sum_{j \neq i} (x_{ij}^k + x_{ji}^k)$. If all trade were intra-industry trade, this index would be equal to zero. Conversely, if all trade were inter-industry trade, it would be equal to 100. In our data, the mean value of this index is 26.9, which resonates well with the well-known fact that the majority of trade among OECD countries is intra-industry when industries are measured at the two-digit level.

The second column of Table 7 reports the change in the previous index averaged across all industries (weighted by total trade in each industry). As expected, the extent of inter-industry trade goes down for most reference countries. Note, however, that (i) “removing a country’s Ricardian comparative advantage” never gets rid of all inter-industry trade and (ii) inter-industry trade may actually go up for some countries after Ricardian comparative advantage has been removed. Two simple explanations for these patterns are the existence of heterogeneous trade

costs and heterogeneous preferences across countries in practice. We will come back to both of these explanations in detail below.

5.3. Welfare

To assess the welfare importance of Ricardian comparative advantage at the industry level, we now compute changes in welfare in the reference country using equation (15). The last two columns of Table 7 present the change in welfare associated with removing Ricardian comparative advantage at the industry level. Seen through the lens of our Ricardian model, eliminating relative industry-level productivity differences across countries leads, on average, to a 0.5% decrease of real income (spent on manufacturing) or only a 5.3% decrease in the overall gains from trade.³³

This small average effect, however, masks a tremendous amount of heterogeneity across countries. Countries such as Australia and Greece appear to be strong beneficiaries of inter-industry Ricardian motives for trade—they each see over a third of their total gains from trade eroded by our counterfactual scenario. For a few countries in our sample, however, the removal of Ricardian comparative advantage actually leads to a welfare gain. Two obvious candidates for explaining these surprising results—heterogeneous trade costs and heterogeneous tastes—were mentioned briefly in Section 5.2. In a standard Ricardian model without trade costs and with identical homothetic preferences across countries, removing differences in relative productivity should always lower trade and welfare. In general, however, removing such differences may very well increase both trade flows and welfare if relative productivity differences are initially correlated with trade costs or differences in preferences. To see this, consider an extreme example in which two countries have not only different relative productivity levels but also different relative demand across goods so that their relative autarky prices are identical. These offsetting effects may be due to endogenous preference formation, as in Atkin (2009), or endogenous technology adoption, as in Acemoglu (2003). Since there is initially no trade between these two countries, it is clear that “removing a country’s Ricardian comparative advantage” would always increase welfare by creating differences in relative autarky prices, and therefore, gains from trade.

A simple way to investigate the quantitative importance of the two previous explanations is to redo our counterfactual exercises under the assumption that expenditure shares do not differ across countries and that trade costs satisfy $d_{ij}^k = d_{ij} \cdot d_j^k$, which, as shown in Corollary 1, implies that trade costs no longer affect the pattern of inter-industry trade.³⁴ The results of our new counterfactuals are reported in Table 8. On average, the welfare impact of Ricardian comparative advantage as a fraction of the total gains from trade (reported in Column (4) of Table 8) goes up from 5.3% in our baseline scenario to (i) 11.7% in the absence of preference differences; (ii) 9.9% in the absence of trade costs violating $d_{ij}^k = d_{ij} \cdot d_j^k$; and (iii) 27.3% in the absence of both. The results of this extension for all countries can be found in the working paper version of this paper, Costinot, Donaldson and Komunjer (2010). As expected, the welfare change only becomes negative for all countries under scenario (iii) since only this scenario assumes a world in which the only source of comparative advantage is relative productivity differences.

33. Normalizing the welfare impact of Ricardian comparative advantage by the total gains from trade allows us to ignore considerations related to the size of the non-tradable sector, which would affect both in a proportional manner. The formula for the total gains from trade is a simple multi-sector extension of the formula used by Eaton and Kortum (2002). It can be found, e.g. in Donaldson (2008).

34. Formally, in the scenarios that follow, in scenario (i), we assume that shares of expenditures α_j^k in country j and industry k are equal to the world share of expenditure in industry k (so that all countries have the same expenditure shares). And in scenario (ii), we assume that trade flows are equal to $\ln x_{ij}^k = \delta_{ij} + \delta_j^k + \delta_i^k$, i.e. we assume that the error term, ε_{ij}^k , in equation (19) exists only due to measurement error in trade flows and should be disregarded in a counterfactual analysis that assumes $d_{ij}^k = d_{ij} \cdot d_j^k$.

TABLE 8
Counterfactual results—robustness

Counterfactual scenario	Outcome variable of interest			
	% change in total exports (1)	Change in index inter-industry trade (2)	% change in welfare (3)	% change in welfare relative to the total gains from trade (4)
Baseline (as in Table 7)	2.94	−5.72	−0.49	−5.32
No preference differences	2.63	−5.29	−1.59	−11.74
No trade costs	3.46	−11.17	−0.74	−9.92
No preference differences or trade costs	3.46	−11.17	−2.83	−27.31

Notes: Results from counterfactual calculations in which, one at a time for each country listed (the “reference country”), every other country in the world is given the reference country’s relative productivity levels across industries, while adjusting each country’s absolute productivity levels in such a way as to hold nominal wages fixed around the world (so as to neutralize terms-of-trade effects). The methodology follows Lemma 4 and Theorem 5. Column (1) reports the resulting proportional change in the total volume of the reference country’s exports. Column (2) reports the resulting change in the reference country’s index of inter-industry trade (given in Section 5.2); this index takes the value of 100 in the case of pure inter-industry trade and zero in the case of pure intra-industry trade. Column (3) reports the resulting proportional change in welfare (equal to real income spent on manufacturing) for the reference country; a negative number indicates a welfare loss. And Column (4) reports this change in welfare as a percentage of the total gains from trade; a negative number indicates a loss (and a value of 100 means that this loss is equal to that of moving to autarky). Unweighted world average effects (averaging across all reference countries) of such counterfactuals are reported. The methodology follows Lemma 4 and Theorem 5. The “No preference differences” scenario calculates world expenditure shares across goods such that these shares are identical across countries. The “No trade costs” scenario calculates trade flows in such a way that idiosyncratic trade cost differences are zero; see footnote 34 for details.

The previous results point towards the importance of trade costs and demand differences for thinking about the consequences of technology diffusion. In a textbook Ricardian model, if China gets access to the U.S.’s technology, then trade flows and welfare will fall in the U.S.; see, *e.g.* Samuelson (2005). By contrast, in a world with trade costs and demand differences across countries, the exact same pattern of technology diffusion may very well increase trade and welfare in both countries. While exploring this possibility further is not the focus of our paper, the results of Table 8 are consistent with this idea.

6. CONCLUDING REMARKS

The Ricardian model has long been perceived as a useful pedagogical tool with, ultimately, little empirical content. Over the last 20 years, the Heckscher–Ohlin model, which emphasizes the role of cross-country differences in factor endowments, has generated a considerable amount of empirical work (see, *e.g.* Bowen, Leamer and Sveikauskas, 1987; Treffer, 1993, 1995; Davis and Weinstein, 2001; Schott, 2004). In contrast, the Ricardian model, which emphasizes productivity differences, has generated almost none.

The main reason for this lack of attention is not the existence of strong beliefs regarding the relative importance of factor endowments and technological considerations. Previous empirical work on the Heckscher–Ohlin model unambiguously shows that technology matters. It derives instead from the obvious mismatch between the real world and the extreme assumptions of the standard two-country Ricardian model. Although the deficiencies of the Ricardian model have not lead to the disappearance of technological considerations from the empirical literature, it has had a strong influence on *how* the relationship between technology and trade has been studied. In the empirical Heckscher–Ohlin–Vanek literature, the factor content of trade remains the main variable of interest.

Building on the seminal work of Eaton and Kortum (2002), the present paper has developed a structural Ricardian model that puts productivity differences at the forefront of the analysis of a central question in international economics: What goods do countries trade? Using this model, we have estimated the impact of productivity differences on the pattern of trade across countries and industries without having to rely on bilateral comparisons inspired by a two-country model, unclear orthogonality conditions, *ad hoc* measures of export performance, or measures of productivity that are systematically biased due to trade-driven selection—all of which are limitations of the existing empirical Ricardian literature.

Using trade and productivity data from 1997, we have estimated the key structural parameter of the model, θ , which governs the elasticity with which increases in observed productivity levels, *ceteris paribus*, lead to increased exports. Our estimate of this elasticity, $\theta = 6.53$, is positive (in agreement with our Ricardian theory), robust to alternative estimation procedures, and lies comfortably within the range of existing estimates in the literature, despite our novel method for obtaining it. Finally, we have used our model to quantify the importance of inter-industry Ricardian forces in today's world economy by measuring the welfare consequences of removing Ricardian comparative advantage at the industry level. According to our estimates, the disappearance of such forces would only lead, on average, to a 5.3% decrease in the total gains from trade.

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Supplementary Data

Supplementary data are available at *Review of Economic Studies* online

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