A.1 Introduction

This web appendix contains additional theoretical derivations and supplementary empirical results for the main paper. In Section A.2, we report additional derivations for our theoretical framework from Section 2 of the paper. In Section A.3, we provide further detail on our structural estimation approach from Section 3 of the paper. In Section A.4, we report further information on the data sources and definitions for our U.S. and Chilean data from Section 4 of the paper.

In Section A.5, we report additional empirical results using our U.S. data that supplement those from Section 5 of the paper. In Section A.6, we replicate all of our empirical results from Section 5 of the paper, but using Chilean data instead of U.S. data. In Section A.7, we show that our theoretical approach allows for unobserved differences in product composition within observed product categories.

A.2 Theoretical Framework Derivations

This section of the web appendix reports additional derivations for Section 2 of the paper. Each subsection has the same name as the corresponding subsection in Section 2 of the paper.

A.2.1 Demand

No further derivations required for Section 2.1 of the paper.

A.2.2 Non-traded Sectors

No further derivations required for Section 2.2 of the paper.

A.2.3 Domestic Versus Foreign Varieties Within Tradable Sectors

No further derivations required for Section 2.3 of the paper.

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A.2.4 Exporter Price Indexes

No further derivations required for Section 2.4 of the paper.

A.2.5 Expenditure Shares

This section of the web appendix reports additional derivations for Section 2.5 of the paper. Corresponding to the firm expenditure share ($S_{ft}^{U}$ in equation (12) in the paper), we can define the share of an individual foreign firm in expenditure on foreign imports within a sector ($S_{fjt}^{F}$) as:

$$S_{fjt}^{F} = \frac{\left( \frac{P_{ft}}{q_{ft}^{F}} \right)^{1 - \sigma_{z}^{F}}}{\sum_{i \in \Omega_{jft}^{E}} \sum_{m \in \Omega_{jft}^{E}} \left( \frac{P_{mt}}{q_{mt}^{F}} \right)^{1 - \sigma_{z}^{F}}}, \quad (A.2.1)$$

where we use “blackboard” font $S_{fjt}^{F}$ for the firm expenditure share to emphasize that this variable is defined as a share of expenditure on foreign firms (since $\Omega_{jft}^{E} \equiv \{ \Omega_{jft}^{I} : i \neq j \}$ in the denominator of equation (A.2.1)).

Similarly, we can define the share of an individual tradable sector in all expenditure on tradable sectors ($S_{jgt}^{T}$) as:

$$S_{jgt}^{T} = \frac{\left( \frac{P_{gjt}}{q_{gjt}^{G}} \right)^{1 - \sigma_{z}^{G}}}{\sum_{k \in \Omega^{T}} \left( \frac{P_{gkt}}{q_{gkt}^{G}} \right)^{1 - \sigma_{z}^{G}}}, \quad (A.2.2)$$

where we use the blackboard font $S_{jgt}^{T}$ and superscript $T$ for the sector expenditure share to signal that this variable is defined across tradable sectors (since $\Omega^{T} \subseteq \Omega^{G}$ in the denominator of equation (A.2.2)).

A.2.6 Log-Linear CES Price Index

No further derivations required for Section 2.6 of the paper.

A.2.7 Entry, Exit and the Unified Price Index

In this section of the web appendix, we report additional derivations for Section 2.7 of the paper. In particular, we derive the expression for the change in the unified price index over time, taking into account entry and exit. Using the shares of expenditure on common goods in equation (15) in the paper, the change in the firm price index between periods $t - 1$ and $t$ ($P_{ft}^{F} / P_{ft-1}^{F}$) can be re-written as:

$$\frac{P_{ft}^{F}}{P_{ft-1}^{F}} = \left( \frac{\lambda_{ft}^{U}}{\lambda_{ft-1}^{U}} \right)^{\frac{1}{\sigma_{z}^{F} - 1}} \left[ \frac{\sum_{u \in \Omega_{jft-1}^{U}} \left( \frac{p_{ut}^{U}}{q_{ut}^{U}} \right)^{1 - \sigma_{z}^{U}}}{\sum_{u \in \Omega_{jft-1}^{U}} \left( \frac{p_{ut-1}^{U}}{q_{ut-1}^{U}} \right)^{1 - \sigma_{z}^{U}}} \right]^{\frac{1}{1 - \sigma_{z}^{F}}}, \quad (A.2.3)$$

where the superscript asterisk indicates that a variable is defined for the common set of varieties. We can also define the share of expenditure on an individual common product in expenditure on all common products within the firm as:

$$S_{ut}^{U*} = \frac{\left( \frac{p_{ut}^{U}}{q_{ut}^{U}} \right)^{1 - \sigma_{z}^{U}}}{\sum_{\ell \in \Omega_{jft-1}^{U}} \left( \frac{p_{\ell t}^{U}}{q_{\ell t}^{U}} \right)^{1 - \sigma_{z}^{U}}} = \frac{\left( \frac{p_{ut}^{U}}{q_{ut}^{U}} \right)^{1 - \sigma_{z}^{U}}}{\left( \frac{p_{ft}^{F}}{p_{ft-1}^{F}} \right)^{1 - \sigma_{z}^{F}}}, \quad (A.2.4)$$
Rearranging this common product expenditure share (A.2.4), taking logarithms, and taking means of both sides of the equation, we obtain the following expression for the log of the common goods firm price index ($P_{ft}^*$):

$$\ln P_{ft}^* = \mathbb{E}_{ft}^{U^*} \left[ p_{ut}^{U^*} \right] - \mathbb{E}_{ft}^{U^*} \left[ \varphi_{ut}^{U^*} \right] + \frac{1}{\sigma_{U^*}^2 - 1} \mathbb{E}_{ft}^{U^*} \left[ S_{ut}^{U^*} \right]$$  \hspace{1cm} (A.2.5)

where $\mathbb{E}_{ft}^{U^*} \left[ \ln p_{ut}^{U^*} \right] \equiv \frac{1}{N_{j_{fj},-1}} \sum_{u \in \Omega_{j_{fj},-1}} \ln \left( p_{ut}^{U^*} \right)$; the superscript $U^*$ indicates that the mean is taken across common products; and the subscripts $f$ and $t$ indicate that this mean varies across firms and over time. Taking logarithms in equation (A.2.3), and using the expression for the common goods firm price index in equation (A.2.5), we obtain equation (16) in the paper.

### A.2.8 Model Inversion

In this section of the web appendix, we report additional derivations for Section 2.8 of the paper. In particular, given the observed data on prices and expenditures for each product $\{P_{ut}^{U^*}, X_{ut}^{U^*}\}$ and the substitution parameters $\{s_{U^*}^g, s_{U^*}^F, s_{U^*}^G\}$, the model is invertible, such that unique values of demand/quality can be recovered from the observed data (up to a normalization or choice of units). We start with the solution for product demand/quality in equation (17) in the paper, reproduced below:

$$\frac{\varphi_{ut}^{U^*}}{M_{ft}^{U^*} \left[ \varphi_{ut}^{U^*} \right]} = \frac{p_{ut}^{U^*}}{M_{ft}^{U^*} \left[ p_{ut}^{U^*} \right]} \left( \frac{S_{ut}^{U^*}}{M_{ft}^{U^*} \left[ S_{ut}^{U^*} \right]} \right)^{\frac{1}{\sigma_{U^*}^2 - 1}},$$  \hspace{1cm} (A.2.6)

where we choose units in which to measure product demand/quality such that its geometric mean across common products within each firm is equal to one:

$$M_{ft}^{U^*} \left[ \varphi_{ut}^{U^*} \right] \equiv \left( \prod_{u \in \Omega_{j_{fj},-1}} \varphi_{ut}^{U^*} \right)^{\frac{1}{N_{j_{fj},-1}}} = 1.$$  \hspace{1cm} (A.2.7)

Having solved for product demand/quality ($\varphi_{ut}^{U^*}$) using equations (A.2.6) and (A.2.7), we use equation (3) in the paper to compute the firm price index, as reproduced below:

$$P_{ft}^F = \left[ \sum_{u \in \Omega_{j_{fj}}^{U^*}} \left( \frac{p_{ut}^{U^*}}{\varphi_{ut}^{U^*}} \right)^{1 - \sigma_{U^*}^2} \right]^{\frac{1}{1 - \sigma_{U^*}^2}}.$$  \hspace{1cm} (A.2.8)

Using this solution for the firm price index ($P_{ft}^F$) from equation (A.2.8), we divide the share of a foreign firm in sectoral imports in equation (A.2.1) by its geometric mean across common foreign firms within that sector to obtain the following solution for demand/quality for each foreign firm:

$$\frac{\varphi_{ft}^{F^*}}{M_{fst}^{F^*} \left[ \varphi_{ft}^{F^*} \right]} = \frac{P_{ft}^F}{M_{fst}^{F^*} \left[ P_{ft}^F \right]} \left( \frac{S_{ft}^{F^*}}{M_{fst}^{F^*} \left[ S_{ft}^{F^*} \right]} \right)^{\frac{1}{\sigma_{F^*}^2 - 1}},$$  \hspace{1cm} (A.2.9)
where we choose units in which to measure firm demand/quality such that its geometric mean across common foreign firms within each sector is equal to one:

\[
M_{Fi}^{F*} = \left( \prod_{i \in \Omega_{Fi}^{t-1}} \prod_{f \in \Omega_{F}^{F_i} \Omega_{Fi}^{t-1}} \phi_{Fi}^{F} \right)^{\frac{1}{N_{Fi}^{t-1}}} = 1. \tag{A.2.10}
\]

Having solved for firm demand/quality (\(\phi_{Fi}^{F}\)) for each foreign firm using equations (A.2.9) and (A.2.10), we use equations (7) and (9) in the paper to compute the sector price index, as reproduced below:

\[
P_{G}^{G} = \left( \mu_{ji}^{G} \right)^{\frac{1}{s_{G}^{1}}} \left[ \sum_{i \in \Omega_{G}^{ji} \Omega_{ji}^{t}} \left( \frac{P_{Fi}^{F}}{\phi_{Fi}^{F}} \right)^{1-s_{G}^{1}} \right]^{1-s_{G}^{1}}, \tag{A.2.11}
\]

where recall that \(\mu_{ji}^{G}\) is the observed share of expenditure on foreign varieties within each sector.

Using this solution for the sector price index (\(P_{G}^{G}\)) from equation (A.2.11), we divide the share of an individual tradable sector in all expenditure on tradable sectors in equation (A.2.2) by its geometric mean across these tradable sectors to obtain the following solution for sector demand for each tradable sector:

\[
\frac{\phi_{ji}^{G}}{M_{ji}^{T}} = \frac{P_{ji}^{G}}{M_{ji}^{T}} \left( \frac{S_{ji}^{T}}{M_{ji}^{T}} \right)^{\frac{1}{1-s_{ji}^{T}}}, \tag{A.2.12}
\]

where we choose units in which to measure sector demand/quality such that its geometric mean across tradable sectors is equal to one:

\[
M_{ji}^{T} = \left( \prod_{g \in \Omega_{ji}^{G}} \phi_{ji}^{G} \right)^{\frac{1}{N_{ji}^{T}}} = 1. \tag{A.2.13}
\]

Recall that there is no asterisk in the superscript of the geometric mean operator across tradable sectors, because the set of tradable sectors is constant over time. Having solved for sector demand/quality (\(\phi_{ji}^{G}\)) for each tradable sector using equations (A.2.12) and (A.2.13), we use equations (4) and (6) in the paper to compute the aggregate price index, as reproduced below:

\[
P_{ji}^{T} = \left( \mu_{ji}^{T} \right)^{\frac{1}{\sigma_{ji}^{1}}} \left[ \sum_{g \in \Omega_{ji}^{T}} \left( P_{ji}^{G} / \phi_{ji}^{G} \right)^{1-\sigma_{ji}^{1}} \right]^{1-\sigma_{ji}^{1}}, \tag{A.2.14}
\]

where recall that \(\mu_{ji}^{T}\) is the observed share of aggregate expenditure on tradable sectors. This completes our inversion of the model to recover the structural residuals for product, firm and sector demand/quality \{\(\phi_{ui}^{U}\), \(\phi_{Fi}^{F}\), \(\phi_{ji}^{G}\)\}.

### A.2.9 Exporter Price Movements

In this section of the web appendix, we report additional derivations for Section 2.9 of the paper. In particular, we derive the log linear decompositions of the exporter price index (\(P_{ji}^{E}\)) for a given exporter and sector in...
equations (18) and (19) in the paper. We first use the CES expression for the share an individual foreign firm \( f \) in country \( j \)'s imports from a foreign exporting country \( i \neq j \) within a sector \( g \):

\[
S_{ji}^{EF} = \frac{\left( P_{fi}^{F} / \varphi_{fit}^{F} \right)^{1-c_g^f} \sum_{k \in \Omega_{jigt}} \left( P_{k}^{F} / \varphi_{kjt}^{F} \right)^{1-c_g^f} \varphi_{kjt}^{F}}{\left( \varphi_{fit}^{F} \right)^{1-c_g^f}}, \quad i \neq j, \tag{A.2.15}
\]

where the superscript \( EF \) is a mnemonic for exporter and firm, and indicates that this firm expenditure share is computed as a share of imports from a single foreign exporting country.

Re-arranging equation (A.2.15), taking logarithms of both sides, adding and subtracting \( \frac{1}{\sigma_g^f-1} \ln N_{jigt}^{F} \), and taking means across foreign firms from that exporter and sector, we obtain the following expression for the log of the exporter price index:

\[
\ln P_{jigt}^{E} = \mathbb{E}_{jigt}^{F} \left[ \ln P_{fit}^{F} \right] - \mathbb{M}_{jigt}^{F} \left[ \ln \varphi_{fit}^{F} \right] - \frac{1}{\sigma_g^f-1} \ln N_{jigt}^{F} + \frac{1}{\sigma_g^f-1} \mathbb{E}_{jigt}^{F} \left[ S_{ji}^{EF} - \ln \frac{1}{N_{jigt}^{F}} \right], \tag{A.2.16}
\]

where \( \mathbb{E}_{jigt}^{F} \left[ \cdot \right] \) is the mean for importer \( j \) across firms from exporter \( i \) within sector \( g \) at time \( t \), such that \( \mathbb{E}_{jigt}^{F} \left[ \ln P_{fit}^{F} \right] \equiv \frac{1}{N_{jigt}^{F}} \sum_{f \in \Omega_{jigt}^{F}} \ln P_{fit}^{F} \).

Substituting the firm price index \( (P_{fit}^{F}) \) from equation (14) in the paper into equation (A.2.16) above, we obtain our exact log linear decomposition of the exporter price index in equation (18) in the paper, which is reproduced below:

\[
\ln P_{jigt}^{E} = \underbrace{\mathbb{E}_{jigt}^{FU} \left[ \ln P_{fit}^{U} \right]}_{(i) \text{ Average log prices}} - \underbrace{\left( \mathbb{E}_{jigt}^{F} \left[ \ln \varphi_{fit}^{F} + \mathbb{E}_{jigt}^{FU} \left[ \ln \varphi_{fit}^{U} \right] \right] \right)}_{(ii) \text{ Average log demand}} + \underbrace{\left( \frac{1}{\sigma_g^f-1} \mathbb{E}_{jigt}^{FU} \left[ \ln S_{ji}^{U} - \ln \frac{1}{N_{jigt}^{F}} \right] + \frac{1}{\sigma_g^f-1} \mathbb{E}_{jigt}^{F} \left[ \ln S_{ji}^{EF} - \ln \frac{1}{N_{jigt}^{F}} \right] \right)}_{(iii) \text{ Dispersion of demand-adjusted prices}}
\]

\[
- \left( \frac{1}{\sigma_g^f-1} \mathbb{E}_{jigt}^{F} \left[ \ln N_{jigt}^{f} \right] + \frac{1}{\sigma_g^f-1} \ln N_{jigt}^{F} \right), \tag{A.2.17}
\]

where \( \mathbb{E}_{jigt}^{FU} \left[ \cdot \right] \) is the mean for importer \( j \) across firms and products from exporter \( i \) within sector \( g \) at time \( t \), such that \( \mathbb{E}_{jigt}^{FU} \left[ \ln P_{fit}^{U} \right] \equiv \frac{1}{N_{jigt}^{U}} \sum_{f \in \Omega_{jigt}^{F}} \frac{1}{N_{jigt}^{U}} \sum_{u \in \Omega_{jigt}^{U}} \ln P_{fit}^{U} \).

We next incorporate the entry and exit of varieties. The log change in the exact CES price index for an importer \( j \) sourcing goods in sector \( g \) from an exporter \( i \) between periods \( t - 1 \) and \( t \) is:

\[
\frac{\Delta P_{jigt}^{E}}{\Delta P_{jigt}^{E}} = \left[ \frac{\sum_{f \in \Omega_{jigt}^{F}} \left( P_{fit}^{F} / \varphi_{fit}^{F} \right)^{1-c_g^f} \varphi_{fit}^{F}}{\sum_{f \in \Omega_{jigt}^{F}} \left( P_{fit}^{F} / \varphi_{fit}^{F} \right)^{1-c_g^f} \varphi_{fit}^{F}} \right]^{\frac{1}{1-c_g^f}}, \tag{A.2.18}
\]

where the entry and exit of firms over time implies that \( \Omega_{jigt}^{F} \neq \Omega_{jigt-1}^{F} \). We define the share of expenditure on common firms \( f \in \Omega_{jigt,t-1}^{F} \) within an exporter and sector in periods \( t \) and \( t - 1 \) as:

\[
\lambda_{jigt}^{F} = \frac{\sum_{f \in \Omega_{jigt,t-1}^{F}} \left( P_{fit}^{F} / \varphi_{fit}^{F} \right)^{1-c_g^f} \varphi_{fit}^{F}}{\sum_{f \in \Omega_{jigt,t-1}^{F}} \left( P_{fit}^{F} / \varphi_{fit}^{F} \right)^{1-c_g^f} \varphi_{fit}^{F}}, \quad \lambda_{jigt-1}^{F} = \frac{\sum_{f \in \Omega_{jigt-1}^{F}} \left( P_{fit-1}^{F} / \varphi_{fit-1}^{F} \right)^{1-c_g^f} \varphi_{fit-1}^{F}}{\sum_{f \in \Omega_{jigt-1}^{F}} \left( P_{fit-1}^{F} / \varphi_{fit-1}^{F} \right)^{1-c_g^f} \varphi_{fit-1}^{F}}. \tag{A.2.19}
\]
Using these definitions from equation (A.2.19), the change in the exporter price index in equation (A.2.18) can be re-written in the following form:

\[
\frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}} = \left( \frac{\lambda^E_{jigt}}{\lambda^E_{jigt-1}} \right) \frac{1}{e^{1 - \sigma_g}} \left[ \sum_{f \in \Omega^E_{jigt-1}} \left( \frac{P^f_{ft}}{\phi^F_{ft}} \right)^{1 - \sigma_g} \right]^{1 - \sigma_g} = \left( \frac{\lambda^E_{jigt}}{\lambda^E_{jigt-1}} \right) \frac{1}{e^{1 - \sigma_g}} \frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}}, \tag{A.2.20}\n\]

where the first term \(\left( \frac{\lambda^E_{jigt}}{\lambda^E_{jigt-1}} \right)^{1 - \sigma_g} \) corrects for the entry and exit of firms; the second term \(\frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}}\) is the change in the exporter price index for common firms; and we again use the superscript asterisk to denote a variable for common varieties. Using this notation, we can also define the share of expenditure on an individual common firm in overall expenditure on common firms for an exporter and sector:

\[
S^E_{ft} = \frac{\sum_{m \in \Omega^E_{jigt-1}} \left( \frac{P^f_{mt}}{\phi^F_{mt}} \right)^{1 - \sigma_g}}{\sum_{f \in \Omega^E_{jigt-1}} \left( \frac{P^f_{ft}}{\phi^F_{ft}} \right)^{1 - \sigma_g}} = \frac{\left( \frac{P^F_{ft}}{\phi^F_{ft}} \right)^{1 - \sigma_g}}{\left( \frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}} \right)^{1 - \sigma_g}}. \tag{A.2.21}\n\]

Rearranging equation (A.2.21) so that the exporter price index for common firms \(\Pi^E_{jigt}\) is on the left-hand side, taking logarithms, and taking means across the set of common firms within an exporter and sector, we obtain:

\[
\ln \Pi^E_{jigt} = \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{P^F_{ft}}{\phi^F_{ft}} \right) \right] - \mathbb{E}^F_{jigt} \left[ \ln \phi^F_{ft} \right] + \frac{1}{\sigma_g - 1} \mathbb{E}^F_{jigt} \left[ \ln S^E_{ft} \right], \tag{A.2.22}\n\]

\(\mathbb{E}^F_{jigt} \left[ \cdot \right]\) is the mean across the common set of firms (superscript \(F^*\)) for a given importer (subscript \(j\)), exporter (subscript \(i\)), sector (subscript \(s\)) and time (subscript \(t\)) such that:

\[
\mathbb{E}^F_{jigt} \left[ \ln \left( \frac{P^F_{ft}}{\phi^F_{ft}} \right) \right] = \frac{1}{N^E_{jigt,t-1}} \sum_{f \in \Omega^E_{jigt,t-1}} \ln \left( \frac{P^F_{ft}}{\phi^F_{ft}} \right). \tag{A.2.23}\n\]

Taking differences over time in equation (A.2.22), we obtain the following expression for the log change in the common goods exporter price index:

\[
\ln \left( \frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}} \right) = \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{P^F_{ft}}{P^F_{ft-1}} \right) \right] - \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{\phi^F_{ft}}{\phi^F_{ft-1}} \right) \right] + \frac{1}{\sigma_g - 1} \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{S^E_{ft}}{S^E_{ft-1}} \right) \right]. \tag{A.2.24}\n\]

We now take logarithms in equation (A.2.20) and use equation (A.2.24) to substitute for \(\frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}}\) and arrive at the following expression for the log change in the overall exporter price index:

\[
\ln \left( \frac{\Pi^E_{jigt}}{\Pi^E_{jigt-1}} \right) = \frac{1}{\sigma_g - 1} \ln \left( \frac{\lambda^E_{jigt}}{\lambda^E_{jigt-1}} \right) + \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{P^F_{ft}}{P^F_{ft-1}} \right) \right] - \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{\phi^F_{ft}}{\phi^F_{ft-1}} \right) \right] + \frac{1}{\sigma_g - 1} \mathbb{E}^F_{jigt} \left[ \ln \left( \frac{S^E_{ft}}{S^E_{ft-1}} \right) \right]. \tag{A.2.25}\n\]

Substituting the expression the change in the firm price index from equation (16) in the paper into equation (A.2.25), we obtain equation (19) in the paper, which is reproduced below:
\[
\Delta \ln P^E_{jigt} = E^{FU*}_{jigt} [\Delta \ln F^I_{ult}] - \left\{ E^{E*}_{jigt} [\Delta \ln \varphi^E_{jigt}] + E^{FU*}_{jigt} [\Delta \ln \varphi^H_{ult}] \right\} \\
(\text{i}) \text{ Average log prices} \quad (\text{ii}) \text{ Average log demand}
\]
\[
+ \left\{ \frac{1}{c^E_{js} - 1} E^{FU*}_{jigt} [\Delta \ln S^U_{ult}] + \frac{1}{c^E_{gs} - 1} E^{FU*}_{jigt} [\Delta \ln S^E_{jigt}] \right\} \\
(\text{iii}) \text{ Dispersion of demand-adjusted prices}
\]
\[
+ \left\{ \frac{1}{c^E_{js} - 1} E^{E*}_{jigt} [\Delta \ln \lambda^U_{jigt}] + \frac{1}{c^E_{gs} - 1} \Delta \ln \lambda^E_{jigt} \right\}.
(\text{iv}) \text{ Variety}
\]

where \( \Delta \) is the difference operator such that \( \Delta \ln P^E_{jigt} \equiv \ln \left( \frac{P^E_{jigt}}{P^E_{jigt-1}} \right) \); \( E^{FU*}_{jigt} [\cdot] \) is a mean, first across common products within firms and then across common firms (superscript \( FU^* \)), for a given importer (subscript \( j \)), exporter (subscript \( i \)), sector (subscript \( g \)) and time period (subscript \( t \)) such that:

\[
E^{FU*}_{jigt} [\Delta \ln \varphi^H_{ult}] = \frac{1}{N^U_{jit}, t-1} \sum_{f \in \Omega^U_{jit, t-1}} \frac{1}{N^{FU^*}_{jit, t-1}} \sum_{u \in \Omega^{FU^*}_{jigt, t-1}} \Delta \ln P^I_{ult}.
\]

Recall that our normalization of product demand in equation (A.2.7) implies \( E^{FU*}_{jigt} [\Delta \ln \varphi^H_{ult}] = 0 \). Therefore the log change in the exporter price index in equation (A.2.26) simplifies to:

\[
\Delta \ln P^E_{jigt} = E^{FU*}_{jigt} [\Delta \ln F^I_{ult}] - \left\{ E^{E*}_{jigt} [\Delta \ln \varphi^E_{jigt}] + E^{FU*}_{jigt} [\Delta \ln \varphi^H_{ult}] \right\} \\
(\text{i}) \text{ Average log prices} \quad (\text{ii}) \text{ Average log demand}
\]
\[
+ \left\{ \frac{1}{c^E_{js} - 1} E^{FU*}_{jigt} [\Delta \ln S^U_{ult}] + \frac{1}{c^E_{gs} - 1} E^{FU*}_{jigt} [\Delta \ln S^E_{jigt}] \right\} \\
(\text{iii}) \text{ Dispersion of demand-adjusted prices}
\]
\[
+ \left\{ \frac{1}{c^E_{js} - 1} E^{E*}_{jigt} [\Delta \ln \lambda^U_{jigt}] + \frac{1}{c^E_{gs} - 1} \Delta \ln \lambda^E_{jigt} \right\}.
(\text{iv}) \text{ Variety}
\]

A.2.10 Patterns of Trade Across Sectors and Countries

A.2.10.1 Revealed Comparative Advantage

In this section of the web appendix, we report the derivation of the results in Section 2.10.1 of the paper. In particular, we derive the decompositions of revealed comparative advantage (RCA) in equations (23) and (24) in the paper. From equation (22) in the paper, log RCA is given by:

\[
\ln (RCA_{jigt}) = \left( 1 - c^E_{jg} \right) \left[ \ln \left( P^E_{jigt} \right) - \frac{1}{N^E_{jgt}} \sum_{k \in \Omega^E_{jgt}} \ln \left( P^E_{jgit} \right) \right] - \frac{1}{N^E_{jgt}} \sum_{k \in \Omega^E_{jgt}} \left( 1 - c^E_{jg} \right) \left[ \ln \left( P^E_{jgit} \right) - \frac{1}{N^T_{jgit}} \sum_{h \in \Omega^T_{jgit}} \ln \left( P^E_{jgh} \right) \right].
\]

where recall that \( \Omega^E_{jgt} \equiv \{ \Omega^E_{jgt} : i \neq j \} \) is the set of foreign exporters that supply importer \( j \) within sector \( g \) at time \( t \); \( N^E_{jgt} = \left| \Omega^E_{jgt} \right| \) is the number of elements in this set; \( \Omega^T_{jgit} \) is the set of tradable sectors that importer \( j \) sources from exporter \( i \) at time \( t \); and \( N^T_{jgit} = \left| \Omega^T_{jgit} \right| \) is the number of elements in this set. Using equation (18) in the paper to substitute for the log exporter price index \( \ln (P^E_{jigt}) \) in equation (A.2.29), we obtain the following exact log-linear decomposition of RCA:
\[
\ln (RCA_{jigt}) = \ln (RCA^P_{jigt}) + \ln (RCA^\varphi_{jigt}) + \ln (RCA^S_{jigt}) + \ln (RCA^N_{jigt}). \tag{A.2.30}
\]

(i) Average log prices
(ii) Average log demand
(iii) Dispersion of demand-adjusted prices
(iv) Variety

The first term in equation (A.2.30) captures average product prices:

\[
\ln (RCA^P_{jigt}) = \left\{ \frac{(1 - \epsilon^P)}{N_{jigt}} \sum_{h \in \Omega^P_{jigt}} [\text{E}^{\text{FL}}_{jigt} [\ln P^U_{ilt}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^P_{jigt}} [\text{E}^{\text{FL}}_{jigt} [\ln P^U_{ilt}]] \right\}.
\tag{A.2.31}
\]

where \( \text{E}^{\text{FL}}_{jigt} [\cdot] \) denotes an average, first across products within firms (superscript \( U \)), and next across firms (superscript \( F \)) supplying importer \( j \) from exporter \( i \) within sector \( g \) at time \( t \) such that:

\[
\text{E}^{\text{FL}}_{jigt} [\Delta \ln P^U_{ilt}] = \frac{1}{N^F_{jigt}} \sum_{f \in \Omega^F_{jigt}} \frac{1}{N^U_{ft}} \sum_{u \in \Omega^U_{ft}} \Delta \ln P^U_{ilt}.
\tag{A.2.32}
\]

The second term in equation (A.2.30) incorporates average firm and product demand:

\[
\ln (RCA^\varphi_{jigt}) = \left\{ \frac{(\epsilon^\varphi - 1)}{N_{jigt}} \sum_{h \in \Omega^\varphi_{jigt}} [\text{E}^{F}_{jigt} [\ln \varphi^F_{ft}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^\varphi_{jigt}} [\text{E}^{F}_{jigt} [\ln \varphi^F_{ft}]] \right\} + \left\{ \frac{(\epsilon^\varphi - 1)}{N_{jigt}} \sum_{h \in \Omega^\varphi_{jigt}} [\text{E}^{F}_{jigt} [\ln \varphi^U_{ult}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^\varphi_{jigt}} [\text{E}^{F}_{jigt} [\ln \varphi^U_{ult}]] \right\},
\tag{A.2.33}
\]

where \( \text{E}^{F}_{jigt} [\cdot] \) denotes an average across firms (superscript \( F \)) supplying importer \( j \) from exporter \( i \) within sector \( g \) at time \( t \) such that:

\[
\text{E}^{F}_{jigt} [\Delta \ln \varphi^F_{ft}] = \frac{1}{N^F_{jigt}} \sum_{f \in \Omega^F_{jigt}} \Delta \ln \varphi^F_{ft}.
\tag{A.2.34}
\]

The third term in equation (A.2.30) reflects the dispersion of firm and product demand-adjusted prices, as reflected in the dispersion of firm and product expenditure shares:

\[
\ln (RCA^S_{jigt}) = \left\{ \frac{\text{E}^{F}_{jigt} [\ln S^F_{ijt}]}{N_{jigt}} - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln S^F_{ijt}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln S^F_{ijt}]] \right\} + \left\{ \frac{\text{E}^{F}_{jigt} [\ln S^U_{uft}]}{N_{jigt}} - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln S^U_{uft}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln S^U_{uft}]] \right\},
\tag{A.2.35}
\]

where \( S^U_{uft} \) is defined in equation (12) in the paper and \( S^F_{ijt} \) is defined in equation (A.2.15) of this web appendix.

The fourth and final term in equation (A.2.30) comprises firm and product variety:

\[
\ln (RCA^N_{jigt}) = \left\{ \frac{\text{E}^{F}_{jigt} [\ln N^F_{jigt}]}{N_{jigt}} - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln N^F_{jigt}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln N^F_{jigt}]] \right\} + \left\{ \frac{\text{E}^{F}_{jigt} [\ln N^U_{uft}]}{N_{jigt}} - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln N^U_{uft}]] - \frac{1}{N_{jigt}} \sum_{h \in \Omega^F_{jigt}} [\text{E}^{F}_{jigt} [\ln N^U_{uft}]] \right\},
\tag{A.2.36}
\]

where

\[ A.2.30 \]

\[ A.2.31 \]

\[ A.2.32 \]

\[ A.2.33 \]

\[ A.2.34 \]

\[ A.2.35 \]

\[ A.2.36 \]
where \( N^E_{jigt} \) is the number of firms that supply importer \( j \) from exporting country \( i \) within sector \( g \) at time \( t \); \( N^E_{jigt} \) is the number of exporting countries that supply importer \( j \) within sector \( g \) at time \( t \); \( N^T_{jigt} \) is the number of tradable sectors in which exporting country \( i \) supplies importer \( j \) at time \( t \); and \( N^U_{jigt} \) is the number of products supplied by firm \( f \) at time \( t \).

Taking logarithms and differencing over time in the definition of RCA in equation (22) in the paper, and using the expression for the change in the log exporter price index from equation (A.2.26) of this web appendix, the log change in revealed comparative advantage (RCA) over time can be written as:

\[
\Delta \ln \left( RCA^*_{jigt} \right) = \Delta \ln \left( \text{RCA}^P_{jigt} \right) + \Delta \ln \left( \text{RCA}^S_{jigt} \right) + \Delta \ln \left( \text{RCA}^\text{disp}^S_{jigt} \right) + \Delta \ln \left( \text{RCA}^\text{var}^S_{jigt} \right),
\]

(A.2.37)

where we compute these log changes for all common exporter-sector pairs with positive trade in both periods, as indicated by the asterisks in the superscripts. The first term in equation (A.2.37) captures average log changes in common product prices:

\[
\Delta \ln \left( \text{RCA}^P_{jigt} \right) = \left\{ \left( 1 - \sigma^F_g \right) \left[ \mathbb{E}^F_{jigt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] - \frac{1}{N^F_{jigt-1}} \sum_{h \in \Omega^F_{jigt-1}} \mathbb{E}^F_{jikt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] \right] - \frac{1}{N^F_{jigt-1}} \sum_{h \in \Omega^F_{jigt-1}} \sigma^F_{h} \left[ \mathbb{E}^F_{jikt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] - \frac{1}{N^F_{jigt-1}} \sum_{h \in \Omega^F_{jigt-1}} \mathbb{E}^F_{jhkt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] \right] \right\},
\]

(A.2.38)

where \( \Omega^F_{jigt,t-1} \) is the set of common foreign exporters that supply importer \( j \) within sector \( g \) in both periods \( t - 1 \) and \( t \); \( N^F_{jigt-1} = |\Omega^F_{jigt,t-1}| \) is the number of elements in this set; \( \Omega^T_{jigt,t-1} \) is the set of tradable sectors that importer \( j \) sources from exporter \( i \) in both periods \( t - 1 \) and \( t \); \( N^T_{jigt} = |\Omega^T_{jigt}| \) is the number of elements in this set; \( \mathbb{E}^F_{jigt} \) denotes an average, first across common products within firms and next across common firms (superscript \( F^* \)), supplying importer \( j \) from exporter \( i \) within sector \( g \) at time \( t \) (as defined in equation (A.2.27)). The second term in equation (A.2.37) incorporates average log changes in common firm and product demand:

\[
\Delta \ln \left( \text{RCA}^S_{jigt} \right) = \left\{ \left( 1 - \sigma^S_g \right) \left[ \mathbb{E}^S_{jigt,t-1} \left[ \Delta \ln \phi^S_{ij} \right] - \frac{1}{N^S_{jigt-1}} \sum_{h \in \Omega^S_{jigt-1}} \mathbb{E}^S_{jikt,t-1} \left[ \Delta \ln \phi^S_{ij} \right] \right] - \frac{1}{N^S_{jigt-1}} \sum_{h \in \Omega^S_{jigt-1}} \sigma^S_{h} \left[ \mathbb{E}^S_{jhkt,t-1} \left[ \Delta \ln \phi^S_{ij} \right] - \frac{1}{N^S_{jigt-1}} \sum_{h \in \Omega^S_{jigt-1}} \mathbb{E}^S_{jhkt,t-1} \left[ \Delta \ln \phi^S_{ij} \right] \right] \right\},
\]

(A.2.39)

where \( \mathbb{E}^S_{jigt} \) denotes an average across common firms (superscript \( S^* \)) supplying importer \( j \) from exporter \( i \) within sector \( g \) at time \( t \) (as defined in equation (A.2.23)). Recall that our normalization of product demand in equation (A.2.7) implies that \( \mathbb{E}^S_{jigt} \left[ \Delta \ln \phi^S_{ij} \right] = 0 \), which in turn implies that this second term simplifies to:

\[
\Delta \ln \left( \text{RCA}^\text{disp}^S_{jigt} \right) = \left\{ \left( 1 - \sigma^F_g \right) \left[ \mathbb{E}^F_{jigt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] - \frac{1}{N^F_{jigt-1}} \sum_{h \in \Omega^F_{jigt-1}} \mathbb{E}^F_{jikt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] \right] - \frac{1}{N^F_{jigt-1}} \sum_{h \in \Omega^F_{jigt-1}} \sigma^F_{h} \left[ \mathbb{E}^F_{jhkt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] - \frac{1}{N^F_{jigt-1}} \sum_{h \in \Omega^F_{jigt-1}} \mathbb{E}^F_{jhkt,t-1} \left[ \Delta \ln \phi^F_{ij} \right] \right] \right\},
\]

(A.2.40)
where, in general, \( \mathbb{E}^F_{jt} \left[ \Delta \ln \varphi_{jt}^F \right] \neq \mathbb{E}^F_{jt} \left[ \Delta \ln \varphi_{jt}^F \right] = 0 \) for an individual exporter \( i \neq j \). The third term in equation (A.2.37) encapsulates the dispersion in demand-adjusted prices across common products and firms:

\[
\ln \left( \text{RCA}^{E*}_{jit} \right) = - \begin{cases} 
\left[ \mathbb{E}^F_{jt} \left[ \Delta \ln S^{E*}_{jt} \right] - \frac{1}{N_{jt}} \sum_{k \in \Omega^{E*}_{jt,t-1}} \mathbb{E}^F_{jk,t-1} \left[ \Delta \ln S^{E*}_{jt} \right] \right] \\
- \frac{1}{N_{jt}} \sum_{k \in \Omega^{E*}_{jt,t-1}} \mathbb{E}^F_{jk,t-1} \left[ \Delta \ln S^{E*}_{jt} \right] + \frac{\sigma^2}{N_{jt}} \left[ \mathbb{E}^F_{jt} \left[ \Delta \ln S^{E*}_{jt} \right] \right]^2 - \frac{1}{N_{jt}} \sum_{k \in \Omega^{E*}_{jt,t-1}} \mathbb{E}^F_{jk,t-1} \left[ \Delta \ln S^{E*}_{jt} \right] \\
- \frac{1}{N_{jt}} \sum_{k \in \Omega^{E*}_{jt,t-1}} \mathbb{E}^F_{jk,t-1} \left[ \Delta \ln S^{E*}_{jt} \right] \end{cases}
\]

, (A.2.41)

where \( S^{E*}_{jt} \) is defined in equation (A.2.4) and \( S^{E*}_{jt} \) is defined in equation (A.2.21). The fourth and final term in equation (A.2.37) corresponds to the entry and exit of products and firms:

\[
\ln \left( \text{RCA}^{E}_{jt} \right) = - \begin{cases} 
\left[ \Delta \ln \lambda^F_{jt} - \frac{1}{N_{jt}} \sum_{k \in \Omega^{F}_{jt,t-1}} \Delta \ln \lambda^F_{kt} \right] \\
- \frac{1}{N_{jt}} \sum_{k \in \Omega^{F}_{jt,t-1}} \Delta \ln \lambda^F_{kt} + \frac{\sigma^2}{N_{jt}} \left[ \Delta \ln \lambda^F_{jt} \right]^2 - \frac{1}{N_{jt}} \sum_{k \in \Omega^{F}_{jt,t-1}} \Delta \ln \lambda^F_{kt} \end{cases}
\]

, (A.2.42)

where \( \lambda^F_{jt} \) is defined in equation (15) in the paper and \( \lambda^F_{jt} \) is defined in equation (A.2.19) of this web appendix.

A.2.10.2 Aggregate Trade

In this section of the web appendix, we report additional derivations for Section 2.10.2 of the paper. In particular, we derive the decomposition of countries’ shares of aggregate imports in equation (25) in the paper. We begin by rewriting the share of an individual exporter in aggregate imports in terms of a share of common imports (supplied in both periods \( t \) and \( t - 1 \)) and entry and exit terms. We have the following accounting identity for the share of an individual exporter in aggregate imports:

\[
S^E_{jit} = \frac{X^E_{jit}}{X^E_{jt}} = \frac{X^{E*}_{jit}}{X^{E*}_{jt}} \frac{X^E_{jit}}{X^{E*}_{jit}}
\]

(A.2.43)

where \( X^E_{jit} \) is country \( j \)’s imports from exporter \( i \neq j \) at time \( t \); \( X^T_{jt} = \) country \( j \)’s total imports from all foreign exporters at time \( t \); \( X^{E*}_{jit} \) is country \( j \)’s imports in common sectors pairs (supplied in both periods \( t - 1 \) and \( t \)) from foreign exporter \( i \neq j \); \( X^{E*}_{jit} \) is country \( j \)’s imports in common exporter-sector pairs (supplied in both periods \( t - 1 \) and \( t \)) from all foreign exporters.

We now define two terms that capture entry and exit of exporter-sector pairs over time. First, we define \( \lambda^E_{jit} \) to be the share of imports in common sectors from an individual foreign exporter \( i \neq j \):

\[
\lambda^E_{jit} \equiv \frac{X^{E*}_{jit}}{X^E_{jit}} = \frac{\sum_{g \in \Omega^T_{jit,t-1}} X^E_{jit} g}{\sum_{g \in \Omega^T_{jit,t-1}} X^E_{jit}}
\]

(A.2.44)

where \( \Omega^T_{jit} \) is the set of traded sectors in which country \( j \) imports from exporter \( i \) at time \( t \) and \( \Omega^T_{jit,t-1} \) is the subset of these sectors that are common (supplied in both periods \( t \) and \( t - 1 \)). Second, we define \( \lambda^T_{jt} \) to be
the share of imports from common exporter-sector pairs in imports from all foreign exporters:

\[ \lambda^T_{jt} \equiv \frac{X^{E*}_{jit}}{X^T_{jt}} = \frac{\sum_{g \in \Omega^E_{jgt-1}} X^E_{jigt}}{\sum_{g \in \Omega^E_{jgt-1}} \sum_{m \in \Omega^E_{jgt-1}} X^E_{jmt}}. \]  

(A.2.45)

where \( \Omega^E_{jgt} \) is the set of foreign exporters \( i \neq j \) from which country \( j \) imports in sector \( g \) at time \( t \) and \( \Omega^E_{jgt,t-1} \) is the subset of these foreign exporters that are common (supplied in both periods \( t \) and \( t-1 \)); \( \Omega^E_{jg} \) is the set of sectors in which country \( j \) imports from foreign exporters at time \( t \); and \( \Omega^E_{jgt,t-1} \) is the subset of these sectors that are common (supplied in both periods \( t \) and \( t-1 \)). Third, we define \( S^E_{jit} \) to be the share of an individual exporter \( i \neq j \) in imports from common exporter-sector pairs:

\[ S^E_{jit} = \frac{X^E_{jit}}{X^{E*}_{jit}} = \frac{\sum_{g \in \Omega^E_{jgt-1}} X^E_{jigt}}{\sum_{g \in \Omega^E_{jgt-1}} \sum_{m \in \Omega^E_{jgt-1}} X^E_{jmt}}. \]  

(A.2.46)

Using equations (A.2.44), (A.2.45) and (A.2.46), we can rewrite the share of an individual foreign exporter \( i \neq j \) in country \( j \) imports from equation (A.2.43) in terms of its share of common imports (\( S^E_{jit} \)), an entry and exit term for that exporter (\( \lambda^E_{jit} \)) and an entry and exit term for imports from all foreign exporters (\( \lambda^T_{jt} \)):

\[ S^E_{jit} = \frac{\lambda^T_{jt}}{\lambda^E_{jit}} S^E_{jit}. \]  

(A.2.47)

Using equation (A.2.46) to substitute for \( S^E_{jit} \) in equation (A.2.47), we obtain:

\[ S^E_{jit} = \frac{\lambda^T_{jt}}{\lambda^E_{jit}} \frac{\sum_{g \in \Omega^E_{jgt-1}} X^E_{jigt}}{\sum_{g \in \Omega^E_{jgt-1}} \sum_{m \in \Omega^E_{jgt-1}} X^E_{jmt}}, \]  

(A.2.48)

which using CES demand can be further re-written as:

\[ S^E_{jit} = \frac{\lambda^T_{jt}}{\lambda^E_{jit}} \frac{\sum_{g \in \Omega^E_{jg}} \left( P^E_{jgt} \right)^{1-\sigma^E_g} X^G_{jgt} \left( P^G_{jgt} \right)^{\sigma^E_g-1}}{\sum_{g \in \Omega^E_{jg}} \sum_{m \in \Omega^E_{jg}} \left( P^E_{jmt} \right)^{1-\sigma^E_g} X^G_{jmt} \left( P^G_{jmt} \right)^{\sigma^E_g-1}}. \]  

(A.2.49)

where \( P^E_{jgt} \) is country \( j \)'s price index for exporter \( i \neq j \) in sector \( g \) at time \( t \); \( X^G_{jgt} \) is country \( j \)'s total expenditure on imports from foreign countries in sector \( g \) at time \( t \); and \( P^G_{jgt} \) is country \( j \)'s import price index for sector \( g \) at time \( t \).

To re-write this expression for an exporter’s share of imports in a log-linear form, we now define two terms for the importance of imports in a given sector from a given exporter, one as a share of common imports across all sectors from that exporter, and the other as a share of common imports across all sectors from all foreign exporters. First, we define importer \( j \)'s expenditure on exporter \( i \neq j \) in sector \( g \) at time \( t \) as a share of expenditure on that exporter across all common sectors as:

\[ Z^E_{jgt} \equiv \frac{X^E_{jgt}}{\sum_{k \in \Omega^E_{jgt-1}} X^E_{jkt}} = \frac{\left( P^E_{jgt} \right)^{1-\sigma^E_g} X^G_{jgt} \left( P^G_{jgt} \right)^{\sigma^E_g-1}}{\sum_{k \in \Omega^E_{jgt-1}} \left( P^E_{jkt} \right)^{1-\sigma^E_g} X^G_{jkt} \left( P^G_{jkt} \right)^{\sigma^E_g-1}}. \]  

(A.2.50)
which can be re-arranged to express the denominator from the right-hand side as follows:

\[
\sum_{k \in \Omega_{jlt-1}} \left( \mathbf{P}^E_{jkt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1} = \frac{\left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jgt} \left( \mathbf{P}^G_{jgt} \right)^{\sigma^E_g - 1}}{Z^E_{jgt}}. \tag{A.2.51}
\]

Taking geometric means across common sectors \( g \in \Omega^E_{jlt-1} \), this becomes:

\[
\sum_{k \in \Omega^E_{jlt-1}} \left( \mathbf{P}^E_{jkt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1} = \frac{\left( \mathbf{M}^T_{jlt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}{\mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}, \tag{A.2.52}
\]

where \( \mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right) \equiv \left( \prod_{g \in \Omega^E_{jlt-1}} \mathbf{P}^E_{jgt} \right)^{1/N^E_{jlt-1}} \) and \( N^E_{jlt-1} \) is the number of common sectors that exporter \( i \) supplies to importer \( j \) between periods \( t-1 \) and \( t \). Second, we define importer \( j \)'s expenditure on exporter \( i \neq j \) in sector \( g \) at time \( t \) as a share of expenditure on common sectors from all foreign exporters as:

\[
\mathbf{y}^E_{jgt} = \frac{X^E_{jkt}}{\sum_{k \in \Omega^E_{jlt-1}} \sum_{m \in \Omega^E_{jlt-1}} X^E_{jkt}} = \frac{\left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}{\sum_{k \in \Omega^E_{jlt-1}} \sum_{m \in \Omega^E_{jlt-1}} \left( \mathbf{P}^E_{jkt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}, \tag{A.2.53}
\]

which can be re-arranged to express the denominator from the right-hand side as follows:

\[
\sum_{k \in \Omega^E_{jlt-1}} \sum_{m \in \Omega^E_{jlt-1}} \left( \mathbf{P}^E_{jkt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1} = \frac{\left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}{\mathbf{y}^E_{jgt}}. \tag{A.2.54}
\]

Taking geometric means across common exporters within each sector and across common sectors, this becomes:

\[
\sum_{k \in \Omega^E_{jlt-1}} \sum_{m \in \Omega^E_{jlt-1}} \left( \mathbf{P}^E_{jkt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1} = \frac{\left( \mathbf{M}^T_{jlt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}{\mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}} \tag{A.2.55}
\]

where \( \mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right) \equiv \left( \prod_{g \in \Omega^E_{jlt-1}} \mathbf{P}^E_{jgt} \right)^{1/N^E_{jlt-1}} \) and \( N^E_{jlt-1} \) is the number of common exporter-sectors for importer \( j \) between periods \( t-1 \) and \( t \).

Using these two measures of the importance of country imports from an individual exporter in a given sector from equations (A.2.52) and (A.2.55), we can re-write the country import share in equation (A.2.49) in the following log-linear form:

\[
\mathbf{S}^E_{jlt} = \frac{\lambda^T_{jlt} \mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}}{\mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}} / \mathbf{M}^T_{jlt} \left( \mathbf{P}^E_{jgt} \right)^{1-\sigma^E_g} X^E_{jkt} \left( \mathbf{P}^G_{jkt} \right)^{\sigma^E_g - 1}. \tag{A.2.56}
\]

Taking logarithms, differencing, and re-arranging terms, we obtain the following log-linear decomposition of a country's share of aggregate imports:

\[
\Delta \ln \mathbf{S}^E_{jlt} = \Delta \ln \left( \frac{\lambda^T_{jlt}}{\lambda^T_{jlt}} \right) + \mathbf{E}^T_{jlt} \left( 1 - \sigma^E_g \right) \left( \Delta \ln \mathbf{P}^E_{jgt} \right) - \mathbf{E}^T_{jlt} \left( 1 - \sigma^E_g \right) \left( \Delta \ln \mathbf{P}^E_{jgt} \right) + \Delta \ln \mathbf{P}^T_{jlt} + \Delta \ln \mathbf{Y}^T_{jlt}. \tag{A.2.57}
\]
where $\mathbb{E}_{ji}^{TE*} \left[ \mathbb{P}_{ji}^{E} \right] \equiv \frac{1}{N_{ji}^{E}} \sum_{g \in \Omega_{ji}^{E}} \sum_{j \in \Omega_{ji}^{E}} \mathbb{P}_{ji}^{E}$. The penultimate term ($\Delta \ln \mathbb{K}_{ji}^{T}$) captures changes in exporter-sector scale, as measured by the change in the extent to which country $j$ sources imports from exporter $i$ in large sectors (sectors with high sectoral import expenditures $X_{ji}^{G}$ and low sectoral import price indexes $P_{ji}^{G}$) relative to its overall imports from all exporters:

$$
\Delta \ln \mathbb{K}_{ji}^{T} \equiv \Delta \ln \left[ \frac{\mathbb{M}_{ji}^{TE} \left[ X_{ji}^{G} \right] \left( \mathbb{M}_{ji}^{TE} \left[ P_{ji}^{G} \right] \right)^{c_i^{j} - 1}}{\mathbb{M}_{ji}^{TE} \left[ X_{ji}^{G} \right] \left( \mathbb{M}_{ji}^{TE} \left[ P_{ji}^{G} \right] \right)^{c_i^{j} - 1}} \right].
$$

(A.2.58)

The final term ($\Delta \ln \mathbb{I}_{ji}^{T}$) captures changes in the sectoral concentration of imports, as measured by changes in the importance of country $j$’s imports from exporter $i$ in sector $g$ as a share of common imports from exporter $i$ ($Z_{ji}^{E}$) relative to its share of aggregate common imports ($Y_{ji}^{E}$):

$$
\Delta \ln \mathbb{I}_{ji}^{T} \equiv \Delta \ln \left[ \frac{\mathbb{M}_{ji}^{TE} \left[ Y_{ji}^{E} \right]}{\mathbb{M}_{ji}^{TE} \left[ Z_{ji}^{E} \right]} \right].
$$

(A.2.59)

This final term corresponds to an exact Jensen’s Inequality correction term that allows us to preserve log linearity in our decompositions of both sectoral and aggregate trade. Using equation (A.2.26) to substitute for the exporter price index ($P_{ji}^{E}$) in equation (A.2.57), we obtain the exact log-linear decomposition of changes in country import shares in equation (25) in the paper, as reproduced below:

$$
\Delta \ln S_{ji}^{E} = -\left\{ \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] - \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] \right\}
$$

$$
+ \left\{ \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] - \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] \right\}
$$

$$
+ \left\{ \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] - \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] \right\}
$$

$$
+ \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right] - \mathbb{E}_{ji}^{TU} \left[ \left( c_i^{j} - 1 \right) \Delta \ln P_{ji}^{U} \right]
$$

$$
- \mathbb{E}_{ji}^{TU} \left[ \Delta \ln S_{ji}^{E} \right] - \mathbb{E}_{ji}^{TU} \left[ \Delta \ln S_{ji}^{E} \right]
$$

$$
- \Delta \ln \left( \frac{S_{ji}^{E} \Delta P_{ji}^{U}}{P_{ji}^{U}} \right)
$$

(A.2.60)

(i) Average log prices,

(ii) Average log product demand,

(iii) Average log firm demand,

(iv) Dispersion product demand-adjusted prices,

(v) Dispersion firm demand-adjusted prices,

(vi) Product Variety,

(vii) Firm Variety,

(viii) Country-Sector Variety

where we have the following definitions:

$$
\mathbb{E}_{ji}^{TU} \left[ \Delta \ln P_{ji}^{U} \right] \equiv \frac{1}{N_{ji}^{U}} \sum_{g \in \Omega_{ji}^{U}} \sum_{j \in \Omega_{ji}^{U}} \frac{1}{N_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \Delta \ln P_{ji}^{U},
$$

(A.2.61)

$$
\mathbb{E}_{ji}^{TU} \left[ \Delta \ln P_{ji}^{U} \right] \equiv \frac{1}{N_{ji}^{U}} \sum_{g \in \Omega_{ji}^{U}} \sum_{j \in \Omega_{ji}^{U}} \frac{1}{N_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \Delta \ln P_{ji}^{U},
$$

(A.2.62)

$$
\mathbb{E}_{ji}^{TU} \left[ \Delta \ln S_{ji}^{E} \right] \equiv \frac{1}{N_{ji}^{U}} \sum_{g \in \Omega_{ji}^{U}} \sum_{j \in \Omega_{ji}^{U}} \frac{1}{N_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \Delta \ln S_{ji}^{E},
$$

(A.2.63)

$$
\mathbb{E}_{ji}^{TU} \left[ \Delta \ln S_{ji}^{E} \right] \equiv \frac{1}{N_{ji}^{U}} \sum_{g \in \Omega_{ji}^{U}} \sum_{j \in \Omega_{ji}^{U}} \frac{1}{N_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \sum_{f \in \Omega_{ji}^{U} - 1} \Delta \ln S_{ji}^{E},
$$

(A.2.64)
\[ E_{jt}^T \left[ \Delta \ln \lambda^F_{jgt} \right] = \frac{1}{N_{jt,t-1}} \sum_{g \in \Omega_{jt,t-1}} \Delta \ln \lambda^F_{jgt} \]  
(A.2.65)

\[ E_{jt}^{TEs} \left[ \Delta \ln \lambda^F_{jgt} \right] = \frac{1}{N_{jt,t-1}} \sum_{g \in \Omega_{jt,t-1}} \sum_{i \in \Omega_{jgt,t-1}} \Delta \ln \lambda^F_{jgt}. \]  
(A.2.66)

### A.2.11 Aggregate Prices

In this section of the web appendix, we report additional derivations for Section 2.11 of the paper. In particular, we derive the decompositions of the aggregate price index \( P_{jt} \) and aggregate import price indexes \( \left( E_{jt}^T \left[ \Delta \ln P^G_{jgt} \right] \right) \) in equations (26) and (27) of the paper respectively.

#### Aggregate Price Index

From equation (4) in the paper, the log aggregate price index \( P_{jt} \) can be written in terms of the share of expenditure on tradable sectors \( \mu_{jt}^T \) and the tradables sector price index \( P_{jt}^T \):

\[ \ln P_{jt} = \frac{1}{\sigma^G - 1} \ln \mu_{jt}^T + \ln P_{jt}^T, \]  
(A.2.67)

Now note that the share of individual tradable sector in expenditure on all tradable sectors is given by:

\[ S_{jgt}^T = \frac{\left( P_{jgt}^G / \varphi_{jgt}^G \right)^{1-\sigma^G}}{\sum_{k \in \Omega^T} \left( P_{jkt}^G / \varphi_{jkt}^G \right)^{1-\sigma^G}} = \frac{\left( P_{jgt}^G / \varphi_{jgt}^G \right)^{1-\sigma^G}}{\left( P_{jt}^T \right)^{1-\sigma^G}}. \]  
(A.2.68)

Rearranging equation (A.2.68), and taking geometric means across tradable sectors, we obtain the following expression for the tradables sector price index \( P_{jt}^T \):

\[ P_{jt}^T = \frac{M_{jt}^T \left[ P_{jgt}^G \right]}{M_{jt}^T \left[ \varphi_{jgt}^G \right]} \left( M_{jt}^T \left[ S_{jgt}^T \right] \right)^{\frac{1}{\sigma^G - 1}}, \]  
(A.2.69)

where \( M_{jt}^T \left[ \cdot \right] \) is the geometric mean across tradable sectors (superscript \( T \)) for a given importer (subscript \( j \)) and time period (subscript \( t \)) such that:

\[ M_{jt}^T \left[ P_{jgt}^G \right] = \left( \prod_{g \in \Omega^T} P_{jgt}^G \right)^{\frac{1}{N^T}}. \]  
(A.2.70)

Substituting this expression for the tradable sector price index from equation (A.2.69) into the aggregate price index in equation (A.2.67), we obtain:

\[ \ln P_{jt} = \frac{1}{\sigma^G - 1} \ln \mu_{jt}^T + E_{jt}^T \left[ \ln P_{jgt}^G \right] - E_{jt}^T \left[ \ln \varphi_{jgt}^G \right] + \frac{1}{\sigma^G - 1} E_{jt}^T \left[ \ln S_{jgt}^T \right], \]  
(A.2.71)

where \( E_{jt}^T \left[ \cdot \right] \) is the mean across tradable sectors (superscript \( T \)) for a given importer (subscript \( j \)) and time period (subscript \( t \)) such that:

\[ E_{jt}^T \left[ P_{jgt}^G \right] = \frac{1}{N^T} \sum_{g \in \Omega^T} \ln P_{jgt}^G. \]  
(A.2.72)
Now using the expression for the sectoral price index from equation (7) in the paper, the aggregate price index in equation (A.2.71) can be written in the following form:

$$\ln P_{jt} = \frac{1}{\sigma_g - 1} \ln \mu_{jt}^T + E_{jt}^T \left[ \ln P_{jt}^G \right] + E_{jt}^T \left[ \frac{1}{\sigma_g - 1} \ln p_{jt}^G \right] - E_{jt}^T \left[ \ln q_{jt}^G \right] + \frac{1}{\sigma_g - 1} E_{jt}^T \left[ \ln S_{jt}^T \right],$$  \hspace{1cm} (A.2.73)

where $P_{jt}^G$ is the sectoral import price index and $\mu_{jt}^G$ is the share of expenditure on foreign varieties within each sector. Taking differences over time, noting that the set of tradable sectors is constant over time, we obtain:

$$\Delta \ln P_{jt} = \frac{1}{\sigma_g - 1} \Delta \ln \mu_{jt}^T + E_{jt}^T \left[ \frac{1}{\sigma_g - 1} \Delta \ln p_{jt}^G \right] + E_{jt}^T \left[ \Delta \ln q_{jt}^G \right] + \frac{1}{\sigma_g - 1} E_{jt}^T \left[ \Delta \ln S_{jt}^T \right],$$  \hspace{1cm} (A.2.74)

which corresponds to equation (26) in the paper.

**Aggregate Import Price Indexes** We next derive the decomposition of the final term in equation (A.2.74) for the average change in sectoral import price indexes ($E_{jt}^T \left[ \Delta \ln P_{jt}^G \right]$) that is reported in equation (27) in the paper. From equation (10) in the paper, the change in the import price index over time can be written as:

$$\frac{P_{jt}^G}{P_{jt-1}^G} = \left[ \frac{\sum_{i \in \Omega_{E_{jt}}^T} \left( P_{jt}^E \right)^{1-\sigma_g}}{\sum_{i \in \Omega_{E_{jt}}^T} \left( P_{jt}^E \right)^{1-\sigma_g}} \right]^{\frac{1}{1-\sigma_g}},$$  \hspace{1cm} (A.2.75)

where the entry and exit of exporters over time implies that $\Omega_{E_{jt}}^T$ is not $\Omega_{E_{jt-1}}^T$. We define the share of expenditure on common foreign exporters $i \in \Omega_{E_{jt}}^T$ that supply importer $j$ within sector $g$ in both periods $t-1$ and $t$ as:

$$\lambda_{E_{jt}}^T = \frac{\sum_{i \in \Omega_{E_{jt}}^T} \left( P_{jt}^E \right)^{1-\sigma_g}}{\sum_{i \in \Omega_{E_{jt}}^T} \left( P_{jt}^E \right)^{1-\sigma_g}}, \hspace{1cm} \lambda_{E_{jt}}^{T-1} = \frac{\sum_{i \in \Omega_{E_{jt-1}}^T} \left( P_{jt-1}^E \right)^{1-\sigma_g}}{\sum_{i \in \Omega_{E_{jt-1}}^T} \left( P_{jt-1}^E \right)^{1-\sigma_g}},$$  \hspace{1cm} (A.2.76)

where $\Omega_{E_{jt-1}}^T$ is the set of common foreign exporters for importer $j$ within sector $g$ and $N_{E_{jt-1}} = \left| \Omega_{E_{jt-1}}^T \right|$ is the number of elements within this set. Using this definition from equation (A.2.76), the change in the import price index in equation (A.2.75) can be re-written in the following form:

$$\frac{P_{jt}^G}{P_{jt-1}^G} = \left( \frac{\lambda_{E_{jt}}^T}{\lambda_{E_{jt}}^{T-1}} \right)^{\frac{1}{1-\sigma_g}} \left[ \frac{\sum_{i \in \Omega_{E_{jt}}^T} \left( P_{jt}^E \right)^{1-\sigma_g}}{\sum_{i \in \Omega_{E_{jt}}^T} \left( P_{jt}^E \right)^{1-\sigma_g}} \right]^{\frac{1}{1-\sigma_g}} = \left( \frac{\lambda_{E_{jt}}^T}{\lambda_{E_{jt}}^{T-1}} \right)^{\frac{1}{1-\sigma_g}} \frac{P_{jt}^G}{P_{jt-1}^G},$$  \hspace{1cm} (A.2.77)

where the first term ($\left( \frac{\lambda_{E_{jt}}^T}{\lambda_{E_{jt}}^{T-1}} \right)^{\frac{1}{1-\sigma_g}}$) corrects for the entry and exit of exporters; the second term ($\frac{P_{jt}^G}{P_{jt-1}^G}$) is the change in the import price index for common exporters; and we again use the superscript asterisk to
denote a variable for common varieties. We can also define the share of expenditure on an individual common exporter in overall expenditure on common exporters as:

$$
S_{jgt}^E = \frac{\left( P_{E_{jigt}}^E \right)^{1-\sigma_g^E}}{\sum_{h \in \Omega_{jgt,t-1}} \left( P_{E_{jigt}}^E \right)^{1-\sigma_g^E}} = \frac{\left( P_{E_{jigt}}^E \right)^{1-\sigma_g^E}}{\left( P_{G_{jigt}}^G \right)^{1-\sigma_g^E}}.
$$

(A.2.78)

Rearranging equation (A.2.78) so that the import price index for common exporters ($P_{jigt}^G$) is on the left-hand side, dividing by the same expression for period $t - 1$, and taking geometric means across the set of common exporters, we have:

$$
\frac{P_{jigt}^G}{P_{jigt-1}^G} = M_{jigt}^E \left[ \frac{P_{jigt}^E}{P_{jigt-1}^E} \right] \left( M_{jigt}^E \left[ \frac{S_{jigt}^E}{S_{jigt-1}^E} \right] \right)^{\frac{1}{\sigma_g^E}}.
$$

(A.2.79)

where $M_{jigt}^E [ \cdot ]$ is the geometric mean across the common set of foreign exporters (superscript $E^*$) for a given importer (subscript $j$), sector (subscript $g$) and time period (subscript $t$) such that:

$$
M_{jigt}^E \left[ P_{jigt}^E \right] = \left( \prod_{i \in \Omega_{jigt,t-1}} P_{jigt}^E \right)^{1/N_{jigt,t-1}}.
$$

(A.2.80)

Combining equations (A.2.77) and (A.2.79), the overall change in the import price index can be written as:

$$
\frac{P_{jigt}^G}{P_{jigt-1}^G} = \left( \frac{\lambda_{jigt}^E}{\lambda_{jigt-1}^E} \right)^{\frac{1}{\sigma_g^E}} M_{jigt}^E \left[ \frac{P_{jigt}^E}{P_{jigt-1}^E} \right] \left( M_{jigt}^E \left[ \frac{S_{jigt}^E}{S_{jigt-1}^E} \right] \right)^{\frac{1}{\sigma_g^E}}.
$$

(A.2.81)

Taking logarithms in equation (A.2.81), we obtain:

$$
\Delta \ln P_{jigt}^G = \frac{1}{\sigma_g^E - 1} \Delta \ln \lambda_{jigt}^E + E_{jigt}^E \left[ \Delta \ln P_{jigt}^E \right] + \frac{1}{\sigma_g^E - 1} E_{jigt}^E \left[ \Delta \ln S_{jigt}^E \right],
$$

(A.2.82)

where $E_{jigt}^E [ \cdot ]$ is the geometric mean across common exporters (superscript $E^*$) for an importer $j$ within sector $g$ at time $t$ such that:

$$
E_{jigt}^E \left[ \Delta \ln P_{jigt}^E \right] = \frac{1}{N_{jigt,t-1}} \sum_{i \in \Omega_{jigt,t-1}} \Delta \ln P_{jigt}^E.
$$

(A.2.83)

We now derive an expression for the average log change in exporter price indexes ($E_{jigt}^E \left[ \Delta \ln P_{jigt}^E \right]$) on the right-hand side of equation (A.2.82). Taking the mean across common exporters in equation (A.2.26), we obtain:

$$
E_{jigt}^E \left[ \Delta \ln P_{jigt}^E \right] = E_{jigt}^{EFUs} \left[ \Delta \ln P_{jigt}^u \right] \left( \frac{1}{\sigma_i^u - 1} \right) + \left( \frac{1}{\sigma_i^u - 1} \right) \cdot E_{jigt}^{EFUs} \left[ \Delta \ln S_{jigt}^u \right] \left( \frac{1}{\sigma_i^u - 1} \right)
$$

(A.2.84)
where $E_{it}^{EFU^t} [\cdot]$ is the mean, first across common products within firms, next across common firms within each exporter-sector, and then across common exporting countries (superscript $EFU^t$), for a given importer (subscript $j$), sector (subscript $g$) and time period (subscript $t$), such that:

$$E_{it}^{EFU^t} \left[ \Delta \ln P_{it}^U \right] = \frac{1}{N_{it}} \sum_{j \in C_{jt}} \frac{1}{N_{it}} \sum_{f \in C_{jt}} \frac{1}{N_{it}} \sum_{u \in C_{jt}} \Delta \ln P_{it}^U,$$

(A.2.85)

recall that $E_{it}^{EF^t} [\cdot]$ is the mean, first across common firms (superscript $F^t$), and next across common exporters (superscript $E^t$) for a given importer (subscript $j$), sector (subscript $g$) and time period (subscript $t$), as defined in equation (A.2.27). Substituting equation (A.2.84) into equation (A.2.26), we obtain the following expression for the change in the sectoral import price index ($\Delta \ln P_{it}^G$) in equation (A.2.82) above:

$$\Delta \ln P_{it}^G = E_{it}^{EFU^t} \left[ \Delta \ln P_{it}^U \right] - \left\{ E_{it}^{EF^t} \left[ \Delta \ln \varphi_{it}^F \right] + E_{it}^{EFU^t} \left[ \Delta \ln \varphi_{it}^U \right] \right\}$$

(A.2.86)

Taking averages across tradable sectors in equation (A.2.86), we obtain the following expression (27) in the paper:

$$E_{it}^{\left[ \Delta \ln P_{it}^G \right]} = E_{it}^{\left[ \Delta \ln P_{it}^U \right]} - E_{it}^{\left[ \Delta \ln \varphi_{it}^F \right]} - E_{it}^{\left[ \Delta \ln \varphi_{it}^U \right]}$$

(i) Average log prices

$$+ E_{it}^{\frac{1}{\sigma_{\varphi}^2} \Delta \ln S_{it}^F} + E_{it}^{\frac{1}{\sigma_{\varphi}^2} \Delta \ln S_{it}^U}$$

(iv) Dispersion country-sector demand-adjusted prices

$$+ E_{it}^{\frac{1}{\sigma_{\varphi}^2} \Delta \ln \lambda_{it}^F} + E_{it}^{\frac{1}{\sigma_{\varphi}^2} \Delta \ln \lambda_{it}^U}$$

(vii) Country - Sector Variety

where the means $E_{it}^{\left[ \cdot \right]}$, $E_{it}^{\left[ \Delta \ln P_{it}^U \right]}$, $E_{it}^{\left[ \Delta \ln \varphi_{it}^F \right]}$ and $E_{it}^{\left[ \Delta \ln \varphi_{it}^U \right]}$ are defined in equations (A.2.65), (A.2.62), (A.2.64), (A.2.66) of this web appendix.

**Interpretation** Together equations (A.2.74) and (A.2.87) (which correspond to equations (26) and (27) in the paper) provide an exact log-linear decomposition of the change in the aggregate cost of living. Each term in these equations has an intuitive interpretation. In the paper, we discuss the interpretation of each term in equation (A.2.74). In this section of the web appendix, we now provide a more detailed discussion of the interpretation of each term in equation (A.2.87).

The first term (i), “Average Prices,” captures changes in the average price of common imported products that are supplied in both periods $t$ and $t - 1$. Other things equal, a fall in these average prices ($E_{it}^{\left[ \Delta \ln P_{it}^U \right]} < 0$) reduces average import price indexes and hence the cost of living. The second and third terms (ii) and (iii) incorporate changes in average firm demand ($\varphi_{it}^F$) across common firms and average product demand ($\varphi_{it}^U$) across common products. Our choice of units for product demand in equation (A.2.7) implies that
the second term for the average log change in demand across common products within each firm is zero:

$$E_{jt}^{TEFU} \left[ \ln q_{ut}^{U} \right] = 0.$$  

Our choice of units for firm demand in equation (A.2.10) implies that the unweighted average log change in demand across common foreign firms within each sector is zero:

$$E_{jt}^{TFU} \left[ \Delta \ln q_{ft}^{F} \right] = 0.$$  

However, the average of firm demand in the third term ($E_{jt}^{TEFU} \left[ \Delta \ln q_{ft}^{F} \right]$) involves first averaging across firms within a given foreign exporter, and then averaging across foreign exporters, which corresponds to a weighted average across firms. Although in principle the weighted and unweighted averages across firms could differ from one another, we find that in practice they take similar values, which implies that the third term is close to zero.

The fourth to sixth terms ((iv)-(vi)) summarize the impact of the dispersion in demand-adjusted prices across common exporter-sector pairs, common firms and common products, respectively. “Country-sector demand-adjusted prices” reflects the fact that consumers are made better off if exporters improve performance in their most successful sectors. For example, consumers are better off if Japanese car makers and Saudi oil drillers become more relatively more productive (raising dispersion in demand-adjusted prices) than if Saudi car makers and Japanese oil drillers are the relative winners (lowering dispersion in demand-adjusted prices). Similarly at the firm-level, consumers benefit more from relative cost reductions or quality improvements for firms with low demand-adjusted prices (high expenditure shares), which increases the dispersion of demand-adjusted prices. Since varieties are substitutes ($\sigma_{s}^{U} > 1$ and $\sigma_{s}^{F} > 1$), increases in the dispersion of these demand-adjusted prices reduce the cost of living, as consumers can substitute away from high-demand-adjusted-price varieties to low-demand-adjusted-price varieties.

The seventh to eighth terms ((vii)-(viii)) summarize the effect of the entry/exit of exporter-sector pairs, firms and products respectively. “Firm Variety” accounts for the entry and exit of foreign firms when at least one foreign firm from an exporter and sector exports in both time periods. “Country-Sector Variety” is an extreme form of foreign firm entry and exit that arises when the number of firms from a foreign exporter rises from zero to a positive value or falls to zero. Finally, the last term (ix), “Product Variety,” accounts for changes in the set of products within continuing foreign firms. For all three terms, the lower the shares of expenditure on common varieties at time $t$ relative to those at time $t - 1$ (the smaller values of $\Delta \ln \lambda_{jg}^{E}$, $\Delta \ln \lambda_{jg}^{F}$ and $\Delta \ln \lambda_{jg}^{U}$), the more attractive are entering varieties relative to exiting varieties, and the greater the reduction in the cost of living between the two time periods.

### A.3 Structural Estimation

In this section of the web appendix, we provide further details on our structural estimation approach from Section 3 of the paper. In particular, we extend the reverse-weighting estimator of Redding and Weinstein (2016) to a nested CES demand system. We exploit the separability properties of CES, which imply that the unit expenditure function can be partitioned into that for a subset of varieties and the expenditure share on this subset of varieties. We use this property to estimate the elasticities of substitution across products, firms and sectors ($\sigma_{s}^{U}$, $\sigma_{s}^{F}$, $\sigma_{g}^{G}$) using only our international trade transactions data. We also use the nesting structure of our model, which implies that the estimation problem is recursive. In a first step, we estimate
the elasticity of substitution across products ($\sigma_{g}^{U}$) for each sector $g$, as discussed in Subsection A.3.1 below. In a second step, we estimate the elasticity of substitution across firms ($\sigma_{g}^{F}$) for each sector $g$, as discussed in Subsection A.3.2 below. In a third step, we estimate the elasticity of substitution across sectors ($\sigma_{g}^{G}$), as discussed in Subsection A.3.3 below. In robustness tests, we also report results using alternative estimates for these elasticities of substitution ($\sigma_{g}^{U}, \sigma_{g}^{F}, \sigma_{g}^{G}$), as discussed in Sections 5.1 and 5.3 of the paper. Additionally, in Subsection A.3.4 below, we report the results of a Monte Carlo simulation, in which we show that our estimation approach successfully recovers the true parameter values when the data are generated according to the model. Finally, in Subsection A.3.5 below, we consider a further robustness check using an alternative representation of the reverse-weighting estimator, which imposes more of the nesting structure of the model.

### A.3.1 Elasticity of Substitution Across Products ($\sigma_{g}^{U}$)

In our first step, we derive three equivalent expressions for the change in the firm price index between periods $t-1$ and $t$ for all foreign firms from exporting countries $i \neq j$. These three expressions use the forward difference of equation (3) in the paper, the backward difference of equation (3) in the paper and our unified price index in equation (16) in the paper. Following analogous steps as in Redding and Weinstein (2016), we obtain the following three equivalent expressions for the change in each firm’s price index between periods $t-1$ and $t$:

\[
\begin{align*}
\frac{P_{ft}}{P_{ft-1}} &= \left( \frac{\lambda_{ft}^{U}}{\lambda_{ft-1}^{U}} \right)^{\frac{1}{1-\sigma_g^{U}}} \left[ \sum_{u \in \Omega_{ft,t-1}^{U}} S_{ut}^{U+} \left( \frac{P_{u}^{U+}/q_{ut}^{U+}}{P_{u}^{U}/q_{ut}^{U}} \right) \right]^{\frac{1}{1-\sigma_g^{U}}}, \\
\frac{P_{ft}}{P_{ft-1}} &= \left( \frac{\lambda_{ft}^{U}}{\lambda_{ft-1}^{U}} \right)^{\frac{1}{1-\sigma_g^{U}}} \left[ \sum_{u \in \Omega_{ft,t-1}^{U}} S_{ut}^{U+} \left( \frac{P_{u}^{U+}/q_{ut}^{U+}}{P_{u}^{U}/q_{ut}^{U}} \right)^{-1(1-\sigma_g^{U})} \right]^{-\frac{1}{1-\sigma_g^{U}}}, \\
\frac{P_{ft}}{P_{ft-1}} &= \left( \frac{\lambda_{ft}^{U}}{\lambda_{ft-1}^{U}} \right)^{\frac{1}{1-\sigma_g^{U}}} \left[ \sum_{u \in \Omega_{ft,t-1}^{U}} S_{ut}^{U+} \left( \frac{P_{u}^{U+}/q_{ut}^{U+}}{P_{u}^{U}/q_{ut}^{U}} \right) \right]^{\frac{1}{1-\sigma_g^{U}}},
\end{align*}
\]

(A.3.1)  

(A.3.2)  

(A.3.3)

where $\Omega_{ft,t-1}^{U}$ is the set of common products that are supplied in both periods $t-1$ and $t$; $\lambda_{ft}^{U}$ and $\lambda_{ft-1}^{U}$ are the expenditure shares on these common products within each firm:

\[
\lambda_{ft}^{U} = \frac{\sum_{u \in \Omega_{ft,t-1}^{U}} \left( \frac{P_{u}}{q_{ut}} \right)^{1-\sigma_g^{U}}}{\sum_{u \in \Omega_{ft}^{U}} \left( \frac{P_{u}}{q_{ut}} \right)^{1-\sigma_g^{U}}}, \quad \lambda_{ft-1}^{U} = \frac{\sum_{u \in \Omega_{ft,t-1}^{U}} \left( \frac{P_{u}}{q_{ut}} \right)^{1-\sigma_g^{U}}}{\sum_{u \in \Omega_{ft-1}^{U}} \left( \frac{P_{u}}{q_{ut}} \right)^{1-\sigma_g^{U}}}.
\]

(A.3.4)

In equations (A.3.1)-(A.3.3), an asterisk denotes the value of a variable for the common set of products ($\Omega_{ft,t-1}^{U}$) such that $S_{ut}^{U*}$ is the share of an individual product in expenditure on all common products within each firm:

\[
S_{ut}^{U*} = \frac{\left( \frac{P_{u}}{q_{ut}} \right)^{1-\sigma_g^{U}}}{\sum_{i \in \Omega_{ft,t-1}^{U}} \left( \frac{P_{u}}{q_{ut}} \right)^{1-\sigma_g^{U}}},
\]

(A.3.5)
\( M^U_{ft} \) is the geometric mean operator across all common products (superscript \( U^* \)) within each firm (subscript \( f \)) and time period (subscript \( t \)) such that:

\[
M^U_{ft} \left[ p^U_{ut} \right] \equiv \left( \prod_{u \in \Omega^U_{ft-1}} p^U_{ut} \right)^{-1_{ft-1}} ;
\]

(A.3.6)

and we choose units in which to measure product demand such that its geometric mean across all common products within each firm is equal to one in each period:

\[
M^U_{ft} \left[ q^U_{ut} \right] \equiv \left( \prod_{u \in \Omega^U_{ft-1}} q^U_{ut} \right)^{-1_{ft-1}} = 1.
\]

(A.3.7)

Recall that we allow for a firm demand/quality shifter \((q^F_{ft})\) that shifts the sales of all products proportionately, which implies that the product demand/quality shifter \((q^U_{ut})\) corresponds to a shock to the relative demand/quality for products within the firm. Using the three equivalent expressions for the change in each firm’s price index in equations (A.3.1)-(A.3.3), and re-arranging terms, we obtain the following two equalities:

\[
\Theta^U_{+ f,t-1} \left[ \sum_{u \in \Omega^U_{ft-1}} S^U_{ut-1} \left( \frac{p^U_{ut}}{p^U_{ut-1}} \right)^{1-\sigma^U_g} \right]^{-1} = M^U_{ft} \left[ \frac{p^U_{ut}}{p^U_{ut-1}} \right] \left( M^U_{ft} \left[ \frac{S^U_{ut}}{S^U_{ut-1}} \right] \right)^{\frac{1}{\sigma^U_g}},
\]

(A.3.8)

\[
\left( \Theta^U_{- f,t-1} \right)^{-1} \left[ \sum_{u \in \Omega^U_{ft-1}} S^U_{ut} \left( \frac{p^U_{ut}}{p^U_{ut-1}} \right)^{-\frac{1}{\sigma^U_g}} \right]^{-\frac{1}{\sigma^U_g}} = M^U_{ft} \left[ \frac{p^U_{ut}}{p^U_{ut-1}} \right] \left( M^U_{ft} \left[ \frac{S^U_{ut}}{S^U_{ut-1}} \right] \right)^{\frac{1}{\sigma^U_g}},
\]

(A.3.9)

where the variety correction terms \((\lambda^U_{ft-1} \left( \frac{\lambda^U_{ft-1}}{1-\sigma^U_g} \right)^{\frac{1}{\sigma^U_g}})\) have cancelled; \(\Theta^U_{+ f,t-1} \) is a forward aggregate demand shifter (where the plus superscript indicates forward); and \(\Theta^U_{- f,t-1} \) is a backward aggregate demand shifter (where the minus superscript indicates backward). These aggregate demand shifters summarize the impact of shocks to the relative demand/quality for individual products on the overall firm price index:

\[
\Theta^U_{+ f,t-1} = \left[ \frac{\sum_{u \in \Omega^U_{ft-1}} S^U_{ut-1} \left( \frac{p^U_{ut}}{p^U_{ut-1}} \right)^{1-\sigma^U_g} \left( \frac{q^U_{ut}}{Q^U_{ft-1}} \right)^{\sigma^U_g-1}} {\sum_{u \in \Omega^U_{ft-1}} S^U_{ut-1} \left( \frac{p^U_{ut}}{p^U_{ut-1}} \right)^{\sigma^U_g}} \right]^{\frac{1}{1-\sigma^U_g}}.
\]

(A.3.10)

\[
\Theta^U_{- f,t-1} = \left[ \frac{\sum_{u \in \Omega^U_{ft-1}} S^U_{ut} \left( \frac{p^U_{ut}}{p^U_{ut-1}} \right)^{-\frac{1}{\sigma^U_g}} \left( \frac{q^U_{ut}}{Q^U_{ft-1}} \right)^{-\frac{\sigma^U_g}{\sigma^U_g}-1}} {\sum_{u \in \Omega^U_{ft-1}} S^U_{ut} \left( \frac{p^U_{ut}}{p^U_{ut-1}} \right)^{-\frac{1}{\sigma^U_g}}} \right]^{\frac{1}{1-\sigma^U_g}}.
\]

(A.3.11)

We make the identifying assumption that the relative demand shocks cancel out across products such that the aggregate demand shifters are both equal to one:

\[
\Theta^U_{+ f,t-1} = \left( \Theta^U_{- f,t-1} \right)^{-1} = 1.
\]

(A.3.12)

Under this identifying assumption, we estimate the elasticity of substitution across firms within products \((\sigma^U_g)\) for each sector \(g\) using the following two sample moment conditions:
\[
m_{Ut}^{U} (\sigma_{U}^{U}) = \left( \ln \left[ \sum_{u \in U_{ft}} S_{U}^{U} \left( \frac{p_{Ut}}{p_{Ut-1}} \right)^{1-\sigma_{U}^{U}} \right] \right) - \ln \left( \left[ M_{U}^{U} \left[ p_{Ut} \right] S_{U}^{U} \left( \frac{p_{Ut}}{p_{Ut-1}} \right)^{1-\sigma_{U}^{U}} \right] \right) = \left( 0 \right). \quad (A.3.13)
\]

We stack these moment conditions for each foreign firm with two or more products and for all time periods within a given sector. We estimate the elasticity of substitution across products within firms (\( \sigma_{U}^{U} \)) using the following generalized method of moments (GMM) estimator:

\[
\hat{\sigma}_{U}^{U} = \arg \min \left\{ m_{Ut}^{U} \left( \sigma_{U}^{U} \right)^{\prime} \times I \times m_{Ut}^{U} \left( \sigma_{U}^{U} \right) \right\}, \quad (A.3.14)
\]

which we refer to as the “reverse-weighting” (RW) estimator.

This reverse-weighting estimator is robust to allowing for common shocks to demand/quality for all of a firm’s products. We capture such common shocks with the firm demand shifter (\( q_{Ut}^{U} \)). If we instead introduced a Hicks-neutral product demand shifter that scales the demand/quality of all of a firm’s products proportionately (such that \( q_{Ut}^{U} = \theta_{ft}^{U} q_{Ut}^{U} \)), this Hicks neutral demand shifter (like the variety correction term) would cancel from both sides of the equalities in equations (A.3.8)-(A.3.9), leaving our estimator of the elasticity of substitution across products (\( \sigma_{U}^{U} \)) unchanged. Therefore, our identifying assumption in equation (A.3.12) allows for such Hicks-neutral shifters, and we only require that changes in the relative demand of products cancel out across products, leaving the firm price index unchanged.

Redding and Weinstein (2016) provide conditions under which our identifying assumption (A.3.12) is satisfied and the RW estimator is consistent. First, this estimator is consistent as the shocks to the relative demand for each product become small (\( q_{Ut}^{U} / q_{Ut-1}^{U} \rightarrow 1 \)). Second, this estimator is also consistent as the number of common products becomes large (\( N_{Ut-1}^{U} \rightarrow \infty \)) if demand shocks are independently and identically distributed. More generally, this identifying assumption holds up to a first-order approximation (\( \Theta_{ft-1}^{U} \approx \left( \Theta_{ft-1}^{U} \right)^{-1} = 1 \)), and the RW estimator can be interpreted as a first-order approximation to the data.

**A.3.2 Elasticity of Substitution Across Firms (\( \sigma_{S}^{U} \))**

Using our estimate of the elasticity of substitution across products (\( \sigma_{U}^{U} \)) from the first step, we can recover the demand shifter for each product (\( q_{Ut}^{U} \)) and compute the firm price index (\( P_{ft}^{U} \)) for all foreign firms:

\[
q_{Ut}^{U} = \frac{p_{Ut}}{M_{ft}^{U} \left[ p_{Ut} \right]} \left( \frac{S_{Ut}^{U}}{M_{ft}^{U} \left[ S_{Ut}^{U} \right]} \right)^{\frac{1}{\sigma_{U}^{U}-1}}, \quad (A.3.15)
\]

\[
P_{ft}^{U} = M_{ft}^{U} \left[ p_{Ut} \right] \left( M_{ft}^{U} \left[ S_{Ut}^{U} \right] \right)^{\frac{1}{\sigma_{U}^{U}-1}}, \quad (A.3.16)
\]

where we have used our choice of units for product demand such that \( M_{ft}^{U} \left[ q_{Ut}^{U} \right] = 1 \).

In our second step, we use these solutions for the firm price index (\( P_{ft}^{U} \)) in three equivalent expressions for the change in the import price index within each sector between periods \( t - 1 \) and \( t \):
\[
\frac{P_{jglt}^G}{P_{jglt-1}^G} = \left( \frac{\lambda_{jgt}^G}{\lambda_{jgt-1}^G} \right)^{\frac{1}{\sigma_g^G}} \left[ \sum_{i \in \Omega_{jglt-1}} \left( \frac{1}{\sum_{f \in \Omega_{jglt-1}} S_{fjt}^* \left( \frac{p_{fjt}^F / \varphi_{fjt}}{p_{fjt-1}^F / \varphi_{fjt-1}} \right)^{1-\sigma_g^F} \right) \right]^{\frac{1}{1-\sigma_g^F}}, \tag{A.3.17}
\]

\[
\frac{P_{jglt}^G}{P_{jglt-1}^G} = \left( \frac{\lambda_{jgt}^G}{\lambda_{jgt-1}^G} \right)^{\frac{1}{\sigma_g^G}} \left[ \sum_{i \in \Omega_{jglt-1}} \sum_{f \in \Omega_{jglt-1}} S_{fjt}^* \left( \frac{p_{fjt}^F / \varphi_{fjt}}{p_{fjt-1}^F / \varphi_{fjt-1}} \right)^{-(1-\sigma_g^F)} \right]^{\frac{1}{1-\sigma_g^F}}, \tag{A.3.18}
\]

\[
\frac{P_{jglt}^G}{P_{jglt-1}^G} = \left( \frac{\lambda_{jgt}^G}{\lambda_{jgt-1}^G} \right)^{\frac{1}{\sigma_g^G}} \frac{M_{jgt}^{F*} \left[ \frac{p_{fjt}^F}{p_{fjt-1}^F} \right]}{M_{jgt}^{F*} \left[ \frac{S_{fjt}^*}{S_{fjt-1}^*} \right]} \left[ \frac{M_{jgt}^{F*} \left[ \frac{S_{fjt}^*}{S_{fjt-1}^*} \right]}{M_{jgt}^{F*} \left[ \frac{S_{fjt}^*}{S_{fjt-1}^*} \right]} \right]^{\frac{1}{1-\sigma_g^F}}, \tag{A.3.19}
\]

where \( \Omega_{jglt-1}^F = \{ \Omega_{jglt-1}^F : i \neq j \} \) is the common set of foreign exporters that supply importer \( j \) within sector \( g \) in both periods \( t-1 \) and \( t \); \( \Omega_{jglt-1}^E \) is the common set of firms that supply importer \( j \) from exporter \( i \) within sector \( g \) in both periods \( t-1 \) and \( t \); \( \lambda_{jgt}^G \) and \( \lambda_{jgt-1}^G \) are the expenditure shares of common foreign firms (supplying in both time periods) in all expenditure on foreign firms within each sector:

\[
\lambda_{jgt}^G = \frac{\sum_{i \in \Omega_{jglt-1}^F} \sum_{f \in \Omega_{jglt-1}^F} \left( \frac{p_{fjt}^F}{\varphi_{fjt}} \right)^{1-\sigma_g^F}}{\sum_{i \in \Omega_{jglt-1}^F} \sum_{f \in \Omega_{jglt-1}^F} \left( \frac{p_{fjt}^F}{\varphi_{fjt}} \right)^{1-\sigma_g^F}}, \quad \lambda_{jgt-1}^G = \frac{\sum_{i \in \Omega_{jglt-1}^F} \sum_{f \in \Omega_{jglt-1}^F} \left( \frac{p_{fjt-1}^F}{\varphi_{fjt-1}} \right)^{1-\sigma_g^F}}{\sum_{i \in \Omega_{jglt-1}^F} \sum_{f \in \Omega_{jglt-1}^F} \left( \frac{p_{fjt-1}^F}{\varphi_{fjt-1}} \right)^{1-\sigma_g^F}}. \tag{A.3.20}
\]

In equations (A.3.17)-(A.3.19), an asterisk denotes the value of a variable for the common set of foreign firms (\( \Omega_{jglt-1}^F \) for \( i \in \Omega_{jglt-1}^F \)) such that \( S_{fjt}^* \) is the share of an individual common foreign firm in expenditure on all common foreign firms within each sector:

\[
S_{fjt}^* = \frac{\left( \frac{p_{fjt}^F}{\varphi_{fjt}} \right)^{1-\sigma_g^F}}{\sum_{i \in \Omega_{jglt-1}^F} \sum_{m \in \Omega_{jglt-1}^F} \left( \frac{p_{fjt-1}^F}{\varphi_{fjt-1}} \right)^{1-\sigma_g^F}}. \tag{A.3.21}
\]

\( M_{jgt}^{F*} \) is the geometric mean operator across all common foreign firms (superscript \( F^* \)) for a given importer (subscript \( j \)), sector (subscript \( g \)) and time period (subscript \( t \)) such that:

\[
M_{jgt}^{F*} \left[ P_{fjt}^F \right] = \left( \prod_{i \in \Omega_{jglt-1}^F} \prod_{f \in \Omega_{jglt-1}^F} p_{fjt}^F \right)^{\frac{1}{N_{jglt-1}^F}}, \tag{A.3.22}
\]

where \( N_{jglt-1}^F \) is the number of common foreign firms serving importer \( j \) in sector \( g \) between periods \( t-1 \) and \( t \); we choose units in which to measure firm demand such that its geometric mean across all common foreign firms within each sector is equal to one:

\[
M_{jgt}^{F*} \left[ \varphi_{fjt}^F \right] = \left( \prod_{i \in \Omega_{jglt-1}^F} \prod_{f \in \Omega_{jglt-1}^F} \varphi_{fjt}^F \right)^{\frac{1}{N_{jglt-1}^F}} = 1. \tag{A.3.23}
\]

Recall that we allow for a sector demand/quality shifter \( (\varphi_{jgt}^G) \) that shifts the sales of all firms proportionately, which implies that the firm demand/quality shifters \( (\varphi_{fjt}^F) \) correspond to a shock to the relative demand/quality
for firms within the sector. Using the three equivalent expressions for the change in the sector import price index in equations (A.3.17)-(A.3.19), we obtain the following two equalities:

$$
\Theta_{\text{fg},t}^{+} = \left[ \sum_{i \in \Omega_{\text{fg},t}^{+}} \sum_{f \in \Omega_{\text{fg},t}^{+}} S_{fi}^{+} \left( \frac{p_{fi}^{+}}{p_{fi}^{+}} \right) \right]^{1-\sigma_{g}^{F}} = M_{\text{fg},t}^{+} \left[ \sum_{i \in \Omega_{\text{fg},t}^{+}} S_{fi}^{+} \right]^{1-\sigma_{g}^{F}} , \tag{A.3.24}
$$

$$
\Theta_{\text{fg},t}^{-} = \left[ \sum_{i \in \Omega_{\text{fg},t}^{-}} \sum_{f \in \Omega_{\text{fg},t}^{-}} S_{fi}^{-} \left( \frac{p_{fi}^{-}}{p_{fi}^{-}} \right) \right]^{1-\sigma_{g}^{F}} = M_{\text{fg},t}^{-} \left[ \sum_{i \in \Omega_{\text{fg},t}^{-}} S_{fi}^{-} \right]^{1-\sigma_{g}^{F}} , \tag{A.3.25}
$$

where the variety correction terms $\left(\frac{\lambda_{G}^{G}}{\lambda_{G}^{G}}\right)^{1/(\sigma_{g}^{F}-1)}$ have cancelled; $\Theta_{\text{fg},t}^{+}$ is a forward aggregate demand shifter (where the plus superscript indicates forward); and $\Theta_{\text{fg},t}^{-}$ is a backward aggregate demand shifter (where the minus superscript indicates backward). These aggregate demand shifters summarize the impact of shocks to demand/quality for individual products on the overall sector import price index:

$$
\Theta_{\text{fg},t}^{+} \equiv \left[ \sum_{i \in \Omega_{\text{fg},t}^{+}} S_{fi}^{+} \left( \frac{p_{fi}^{+}}{p_{fi}^{+}} \right) \right]^{1-\sigma_{g}^{F}} , \tag{A.3.26}
$$

$$
\Theta_{\text{fg},t}^{-} \equiv \left[ \sum_{i \in \Omega_{\text{fg},t}^{-}} S_{fi}^{-} \left( \frac{p_{fi}^{-}}{p_{fi}^{-}} \right) \right]^{1-\sigma_{g}^{F}} . \tag{A.3.27}
$$

We make the identifying assumption that the relative demand shocks cancel out across firms such that the aggregate demand shifters are both equal to one:

$$
\Theta_{\text{fg},t}^{+} = \Theta_{\text{fg},t}^{-} = 1 , \tag{A.3.28}
$$

Under this identifying assumption, we estimate the elasticity of substitution across firms ($\sigma_{g}^{F}$) for each sector $g$ using the following two sample moment conditions:

$$
\begin{align*}
& m_{g}^{F} (\sigma_{g}^{F}) = \\
& \left( \ln \left\{ \sum_{i \in \Omega_{\text{fg},t}^{+}} S_{fi}^{+} \left( \frac{p_{fi}^{+}}{p_{fi}^{+}} \right) \right\}^{1-\sigma_{g}^{F}} - \ln \left\{ \sum_{i \in \Omega_{\text{fg},t}^{-}} S_{fi}^{-} \left( \frac{p_{fi}^{-}}{p_{fi}^{-}} \right) \right\}^{1-\sigma_{g}^{F}} \right) \times M_{g}^{F} \left( \sigma_{g}^{F} \right) + r_{g}^{F} \left( \sigma_{g}^{F} \right) + M_{g}^{F} \left( \sigma_{g}^{F} \right) + r_{g}^{F} \left( \sigma_{g}^{F} \right) = 0 . \tag{A.3.29}
\end{align*}
$$

We stack these moment conditions for all time periods for a given sector. We estimate the elasticity of substitution across firms ($\sigma_{g}^{F}$) using the following generalized method of moments (GMM) estimator:

$$
\hat{\sigma}_{g}^{F} = \arg \min \left\{ m_{g}^{F} (\sigma_{g}^{F})' \times I \times m_{g}^{F} (\sigma_{g}^{F}) \right\} , \tag{A.3.30}
$$

which we again refer to as the “reverse-weighting” (RW) estimator.

This reverse-weighting estimator is robust to allowing for common shocks shocks to demand/quality for all firms within a sector. We capture such common shocks with the sector demand shifter $\phi_{\text{fg},t}^{G}$. If we were to instead introduce a Hicks-neutral firm demand shifter that scales the demand/quality of all firms within a
sector proportionately (such that \( \varphi_{jt}^F = \theta_{jt}^G \varphi_{jt}^F \)), this Hicks neutral demand shifter (like the variety correction term) would cancel from both sides of the equalities in equations (A.3.24)-(A.3.25), leaving our estimator of the elasticity of substitution across firms \((\sigma_{jt}^F)\) unchanged. Therefore, our identifying assumption in equation (A.3.28) allows for such Hicks-neutral shifters, and we only require that changes in the relative demand of firms cancel out across firms, leaving the sector import price index unchanged.

As discussed above, Redding and Weinstein (2016) provide conditions under which our identifying assumption (A.3.28) is satisfied and the RW estimator is consistent. First, this estimator is consistent as the number of common shocks to the relative demand for each firm become small \((\varphi_{jt}^F / \varphi_{jt-1}^F \to 1)\). Second, this estimator is also consistent as the number of common firms becomes large \((N_{jt}^F \to \infty)\) if demand shocks are independently and identically distributed. More generally, this identifying assumption holds up to a first-order approximation \((\Theta_{jt}^{F+} / \Theta_{jt-1}^{F-})^{-1} \approx 1\), and the RW estimator can be interpreted as a first-order approximation to the data.

### A.3.3 Elasticity of Substitution Across Sectors \((\sigma^G)\)

Using our estimate of the elasticity of substitution across firms \((\sigma_{jt}^F)\) from the first step, we can recover the demand shifter for each foreign firm \((\varphi_{jt}^F)\) and compute the sector import price index \((P_{jt}^G)\). Combining this solution for the sector import price index \((P_{jt}^G)\) with the share of expenditure within each sector on foreign varieties \((\mu_{jt}^{G_i})\), we can also compute the overall sector price index \((P_{jt}^F)\). In our third step, we use these solutions for the overall sector price index \((P_{jt}^G)\) in three equivalent expressions for the change in the price index across all tradable sectors \(((P_{jt}^T)^{-1})\) between periods \(t - 1\) and \(t\):

\[
\frac{P_{jt}^T}{P_{jt-1}^T} = \left[ \sum_{S \in \Omega^T} S^T_{jt} \left( \frac{P_{jt}^G / \varphi_{jt}^G}{P_{jt-1}^G / \varphi_{jt-1}^G} \right) \right]^{1 - \omega^G} \left( \frac{1}{1 - \omega^G} \right),
\]

(A.3.31)

\[
\frac{P_{jt}^T}{P_{jt-1}^T} = \left[ \sum_{S \in \Omega^T} S^T_{jt} \left( \frac{P_{jt}^F / \varphi_{jt}^F}{P_{jt-1}^F / \varphi_{jt-1}^F} \right) \right]^{1 - \omega^G} \left( \frac{1}{1 - \omega^G} \right),
\]

(A.3.32)

\[
\frac{P_{jt}^T}{P_{jt-1}^T} = M^T_{jt} \left[ \frac{P_{jt}^G}{P_{jt-1}^G} \right] \left( M^T_{jt} \left[ \frac{S^T_{jt}}{S_{jt-1}^T} \right] \right)^{\omega^G - 1},
\]

(A.3.33)

where we have used the fact that the set of tradable sectors \((\Omega^T)\) is constant over time in our data; hence there are no asterisks on variables and no variety correction terms for the entry and exit of sectors; \(S^T_{jt}\) is the share of an individual tradable sector in all expenditure on tradable sectors such that:

\[
S^T_{jt} = \frac{\left( \frac{P_{jt}^G}{\varphi_{jt}^G} \right)^{1 - \omega^G}}{\sum_{k \in \Omega^T} \left( \frac{P_{jt}^G}{\varphi_{jt}^G} \right)^{1 - \omega^G}}.
\]

(A.3.34)

\(M^T_{jt} [\cdot]\) is the geometric mean operator across tradable sectors (superscript \(T\)) for a given importer (subscript
following two sample moment conditions:

\[ \prod_{j \in \Omega_f} P_{jgt}^{G} \right) ^{1/N_T} \]

and we choose units in which to measure sector demand such that its geometric mean across tradable sectors is equal to one:

\[ \prod_{j \in \Omega_f} \varphi_{jgt}^{G} \right) ^{1/N_T} = 1. \] (A.3.36)

Using the three equivalent expressions for the change in the price index for tradable sectors in equations (A.3.31)-(A.3.33), and using the relationship between the overall sector price index and the sector import price index in equation (4) in the paper, we obtain the following two equalities:

\[ \left( P_{jgt}^{G} / P_{jgt}^{G-1} \right)^{(1-\sigma^G)} \right) ^{1/N_T} = M_{jt}^T \left( P_{jgt}^{G} / P_{jgt}^{G-1} \right)^{(1-\sigma^G)} \right) ^{1/N_T}, \] (A.3.37)

\[ \left( P_{jgt}^{G} / P_{jgt}^{G-1} \right)^{(1-\sigma^G)} \right) ^{1/N_T} = M_{jt}^T \left( P_{jgt}^{G} / P_{jgt}^{G-1} \right)^{(1-\sigma^G)} \right) ^{1/N_T}, \] (A.3.38)

where recall that \( P_{jgt}^{G} \) is the sector import price index and \( \mu_{jgt}^{G} \) is the share of expenditure on foreign varieties within each sector; \( \Theta_{jt,t-1}^{T+} \) is a forward aggregate demand shifter (where the plus superscript indicates forward); and \( \Theta_{jt,t-1}^{T-} \) is a backward aggregate demand shifter (where the minus superscript indicates backward). These aggregate demand shifters summarize the impact of shocks to demand/quality for individual tradable sectors on the overall price index for all tradable sectors:

\[ \Theta_{jt,t-1}^{T+} \equiv \left[ \frac{\sum_{j \in \Omega_f} S_{jgt}^{T} \left( \frac{\mu_{jgt}^{G}}{\mu_{jgt}^{G-1}} \right)^{\frac{1}{\sigma^G}} \left( \frac{P_{jgt}^{G}}{P_{jgt}^{G-1}} \right)^{1-\sigma^G} \left( \varphi_{jgt}^{G} / \varphi_{jgt}^{G-1} \right)^{(1\sigma^G)} \right]}{\sum_{j \in \Omega_f} S_{jgt}^{T} \left( \frac{\mu_{jgt}^{G}}{\mu_{jgt}^{G-1}} \right)^{\frac{1}{\sigma^G}} \left( \frac{P_{jgt}^{G}}{P_{jgt}^{G-1}} \right)^{1-\sigma^G}} \right] ^{1/N_T}, \] (A.3.39)

\[ \Theta_{jt,t-1}^{T-} \equiv \left[ \frac{\sum_{j \in \Omega_f} S_{jgt}^{T} \left( \frac{\mu_{jgt}^{G}}{\mu_{jgt}^{G-1}} \right)^{\frac{1}{\sigma^G}} \left( \frac{P_{jgt}^{G}}{P_{jgt}^{G-1}} \right)^{(1-1\sigma^G)} \left( \varphi_{jgt}^{G} / \varphi_{jgt}^{G-1} \right)^{(1-1\sigma^G)} \right]}{\sum_{j \in \Omega_f} S_{jgt}^{T} \left( \frac{\mu_{jgt}^{G}}{\mu_{jgt}^{G-1}} \right)^{\frac{1}{\sigma^G}} \left( \frac{P_{jgt}^{G}}{P_{jgt}^{G-1}} \right)^{(1-1\sigma^G)}} \right] ^{1/N_T}. \] (A.3.40)

We make the identifying assumption that the demand shocks cancel out across sectors such that the aggregate demand shifters are both equal to one:

\[ \Theta_{jt,t-1}^{T+} = \left( \Theta_{jt,t-1}^{T-} \right)^{-1} = 1, \] (A.3.41)

Under this identifying assumption, we estimate the elasticity of substitution across sectors (\( \sigma^G \)) using the following two sample moment conditions:

}\]
In this section, we report the results of a Monte Carlo simulation for our estimation procedure. For simplicity, we focus on the case of a single tradable sector with CES demand defined over two nests for firms and products. We assume a conventional supply-side with monopolistic competition and constant marginal costs (as in Krugman 1980 and Melitz 2003). We first assume true values for the model’s parameters (the elasticities of substitution, $\sigma^F$ and $\sigma^U$) and its structural residuals (firm demand $q^F_{jt}$, product demand $q^U_{it}$, and product marginal cost $\delta^U_{il}$). We next solve for equilibrium prices and expenditure shares. Finally, we suppose that a researcher only observes data on these prices and expenditure shares and implements our reverse-weighting estimation procedure. For each combination of parameters, we undertake 250 replications of the model. We compare the distribution of estimates across these replications with the true parameter values.

As the reverse-weighting estimator uses only the subset of common goods, we focus on this subset, and are not required to make assumptions about entering and exiting goods. We assume 1,000 firms, 1,000 products for each firm, and 2 time periods. We assume values for the true elasticities of substitution across products and firms of four and two respectively ($\sigma^U = 4$ and $\sigma^F = 2$). We draw time-varying values for product demand

\[\hat{\sigma}^G = \text{arg min} \left\{ m^T \left( \sigma^G \right)' \times \mathbb{1} \times m^T \left( \sigma^G \right) \right\}, \quad (A.3.43)\]

This reverse-weighting estimator is robust to allowing for common shocks shocks to demand/quality for all sectors. If we were to introduce a Hicks-neutral sector demand shifter that scales the demand/quality of all sectors proportionately (such that $q^G_{jt} = \theta_j q^G_{jt-1}$), this Hicks neutral demand shifter (like the variety correction term) would cancel from both sides of the equalities in equations (A.3.37)-(A.3.38), leaving our estimator of the elasticity of substitution across sectors ($\sigma^G$) unchanged.

As discussed above, Redding and Weinstein (2016) provide conditions under which our identifying assumption (A.3.41) is satisfied and the RW estimator is consistent. First, this estimator is consistent as the demand shocks for each sector become small ($q^G_{jt}/q^G_{jt-1} \rightarrow 1$). Second, this estimator is also consistent as the number of tradable sectors becomes large ($N^T \rightarrow \infty$) if demand shocks are independently and identically distributed. More generally, this identifying assumption holds up to a first-order approximation ($\Theta^T_{jt-1} \approx \left( \Theta^T_{jt-1} \right)^{-1} \approx 1$), and the RW estimator can be interpreted as a first-order approximation to the data.

A.3.4 Monte Carlo

In this section, we report the results of a Monte Carlo simulation for our estimation procedure. For simplicity, we focus on the case of a single tradable sector with CES demand defined over two nests for firms and products. We assume a conventional supply-side with monopolistic competition and constant marginal costs (as in Krugman 1980 and Melitz 2003). We first assume true values for the model’s parameters (the elasticities of substitution, $\sigma^F$ and $\sigma^U$) and its structural residuals (firm demand $q^F_{jt}$, product demand $q^U_{it}$, and product marginal cost $\delta^U_{il}$). We next solve for equilibrium prices and expenditure shares. Finally, we suppose that a researcher only observes data on these prices and expenditure shares and implements our reverse-weighting estimation procedure. For each combination of parameters, we undertake 250 replications of the model. We compare the distribution of estimates across these replications with the true parameter values.

As the reverse-weighting estimator uses only the subset of common goods, we focus on this subset, and are not required to make assumptions about entering and exiting goods. We assume 1,000 firms, 1,000 products for each firm, and 2 time periods. We assume values for the true elasticities of substitution across products and firms of four and two respectively ($\sigma^U = 4$ and $\sigma^F = 2$). We draw time-varying values for product demand

\[\hat{\sigma}^G = \text{arg min} \left\{ m^T \left( \sigma^G \right)' \times \mathbb{1} \times m^T \left( \sigma^G \right) \right\}, \quad (A.3.43)\]
(\(q^{U}_{ut}\)), firm demand (\(q^{F}_{ft}\)) and marginal cost (\(a^{U}_{ut}\)) from independent log normal distributions. We use these realizations and the equilibrium conditions of the model to solve for product prices (\(P^{U}_{ut}\)), product expenditure shares (\(S^{U}_{ut}\)), firm price indexes (\(P^{F}_{ft}\)) and firm expenditure shares (\(S^{F}_{ft}\)):

\[
P^{U}_{ut} = \frac{\sigma^{U}}{\sigma^{U} - 1} a^{U}_{ut},
\]

(A.3.44)

\[
S^{U}_{ut} = \frac{\left(\frac{P^{U}_{ut}}{q^{U}_{ut}}\right)^{1-\sigma^{U}}}{\sum_{t} \left(\frac{P^{U}_{ut}}{q^{U}_{ut}}\right)^{1-\sigma^{U}}},
\]

(A.3.45)

\[
P^{F}_{ft} = \left[ \sum_{u} \left(\frac{P^{U}_{ut}}{q^{U}_{ut}}\right)^{1-\sigma^{U}} \right]^{\frac{1}{1-\sigma^{U}}},
\]

(A.3.46)

\[
S^{F}_{ft} = \frac{\left(\frac{P^{F}_{ft}}{q^{F}_{ft}}\right)^{1-\sigma^{F}}}{\sum_{m} \left(\frac{P^{F}_{mt}}{q^{F}_{mt}}\right)^{1-\sigma^{F}}},
\]

(A.3.47)

Treating the solutions for product prices (\(P^{U}_{ut}\)), product expenditure shares (\(S^{U}_{ut}\)) and firm expenditure shares (\(S^{F}_{ft}\)) as data, we first estimate the elasticity of substitution across products (\(\sigma^{U}\)) using step one of our estimation procedure outlined above. We nest use this estimate (\(\hat{\sigma}^{U}\)) to solve for product demand (\(\hat{q}^{U}_{ut}\)) and construct the firm price indexes (\(\hat{P}^{F}_{ft}\)). Using these solutions for the firm price indexes (\(\hat{P}^{F}_{ft}\)) and the data on firm expenditure shares (\(S^{F}_{ft}\)), we next estimate the elasticity of substitution across firms (\(\sigma^{F}\)) using step two of our estimation procedure outlined above.

In Figures A.3.1 and A.3.2, we show histograms of the parameter estimates across replications (blue bars) and the true parameter values (red solid vertical line). We find that the mean estimates of both the product elasticity (\(\hat{\sigma}^{U}\)) and the firm elasticity (\(\hat{\sigma}^{F}\)) lie close to the true parameter values. We find somewhat larger dispersion in the firm elasticity (\(\hat{\sigma}^{F}\)) than in the product elasticity (\(\hat{\sigma}^{U}\)), consistent with the firm variables being constructed from the product estimates, which introduces estimation error. In both cases, we are unable to reject the null hypothesis that the estimated parameters are equal to their true values at conventional levels of significance.

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Figure A.3.1: Estimated Elasticity of Substitution Across Products Within Firms ($\hat{s}^U$)

Figure A.3.2: Estimated Elasticity of Substitution Across Firms ($\hat{s}^F$)
A.3.5 Robustness

As a robustness check, we now show that there is an alternative representation of the reverse-weighting estimator for tiers of utility above the lower tier (i.e. for the firm and sector tiers above the product tier). For brevity, we derive this alternative representation for the elasticity of substitution across firms ($\sigma^F_g$), but the same derivation goes through for the elasticity of substitution across sectors ($\sigma^G$). We begin with the expressions for expenditure on each product ($X^U_{ut}$) and expenditure on each firm ($X^F_{ft}$) from CES demand:

\[
X^U_{ut} = \left( \frac{p^U_{ut}}{q^U_{ut}} \right)^{1-\sigma^U_g} X^F_{ft} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{\sigma^U_g-1}, \tag{A.3.48}
\]

\[
X^F_{ft} = \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{1-\sigma^F_g} X^G_{jgt} \left( \frac{p^G_{jgt}}{q^G_{jgt}} \right)^{\sigma^F_g-1}, \tag{A.3.49}
\]

where $X^G_{jgt}$ is importer $j$’s total expenditure on foreign varieties from exporters $i \neq j$ in sector $g$ at time $t$, and $p^G_{jgt}$ is importer $j$’s sectoral import price index for sector $g$ at time $t$. Combining equations (A.3.48) and (A.3.49), we obtain:

\[
X^U_{ut} = \left( \frac{p^U_{ut}}{q^U_{ut}} \right)^{1-\sigma^U_g} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{1-\sigma^F_g} X^G_{jgt} \left( \frac{p^G_{jgt}}{q^G_{jgt}} \right)^{\sigma^F_g-1}, \tag{A.3.50}
\]

\[
X^U_{ut} = \left( \frac{p^U_{ut}}{q^U_{ut}} \right)^{1-\sigma^U_g} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{\sigma^U_g-1} X^G_{jgt} \left( \frac{p^G_{jgt}}{q^G_{jgt}} \right)^{\sigma^F_g-1}, \tag{A.3.51}
\]

Rearranging equation (A.3.51), we get:

\[
\left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{\sigma^U_g-1} = \frac{X^U_{ut}}{X^G_{jgt}} \left( \frac{p^U_{ut}}{q^U_{ut}} \right)^{\sigma^U_g-1} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{-\sigma^F_g-1}, \tag{A.3.52}
\]

\[
\left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{\sigma^U_g-1} = \frac{X^U_{ut}}{X^G_{jgt}} \left( \frac{p^U_{ut}}{q^U_{ut}} \right)^{-\sigma^U_g-1} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{\sigma^F_g-1}, \tag{A.3.53}
\]

Taking geometric means across common products within the firm in equation (A.3.53), we obtain:

\[
p^F_{ft} = \left( \mathbf{M}^U_{ft} \left[ X^U_{ut} / X^G_{jgt} \right] \right)^{\sigma^U_g-1} \left( \mathbf{M}^F_{ft} \left[ P^U_{ut} / P^G_{jgt} \right] \right)^{\sigma^U_g-1} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{-\sigma^F_g-1} \left( \frac{p^G_{jgt}}{q^G_{jgt}} \right)^{-\sigma^F_g-1}, \tag{A.3.54}
\]

where we have used our normalization that $\mathbf{M}^U_{ft} \left[ q^U_{ut} \right] = 1$. Now taking geometric means across common foreign firms within a sector in equation (A.3.54), we have:

\[
\mathbf{M}^G_{jgt} \left[ p^F_{ft} \right] = \left( \mathbf{M}^F_{ft} \left[ X^U_{ut} / X^G_{jgt} \right] \right)^{\sigma^U_g-1} \left( \mathbf{M}^F_{ft} \left[ P^U_{ut} / P^G_{jgt} \right] \right)^{\sigma^U_g-1} \left( \frac{p^F_{ft}}{q^F_{ft}} \right)^{\sigma^F_g-1} \left( \frac{p^G_{jgt}}{q^G_{jgt}} \right)^{-\sigma^F_g-1}, \tag{A.3.55}
\]

where we have used our choice of units that $\mathbf{M}^G_{jgt} \left[ q^F_{ft} \right] = 1$. Finally, using equation (A.3.55) to substitute for $\mathbf{M}^G_{jgt} \left[ p^F_{ft} \right]$ in equation (A.3.19) above, we obtain another equivalent expression for our unified price index that exploits more of the nesting structure of the model:
\[
\frac{p_{gt}^F}{p_{gt-1}^F} = \left( \frac{\lambda_{ft}^{U_L}}{\lambda_{ft-1}^{U_L}} \right)^{\frac{1}{\sigma_g^F - 1}} \left( M_{ft}^{U_L} \left\{ \frac{x_{ft}^{U_L} / x_{ft-1}^{U_L}}{x_{ft-1}^{U_L} / x_{ft-1}^{U_L}} \right\} \right)^{\frac{1}{\sigma_g^F - 1} - 1} \left( M_{ft}^{U_L} \left\{ \frac{p_{lt}^U / p_{lt-1}^U}{p_{lt-1}^U / p_{lt-1}^U} \right\} \right)^{\frac{1}{\sigma_g^F - 1} - 1} \left( M_{ft}^{U_L} \left\{ \frac{s_{ft}^U / s_{ft-1}^U}{s_{ft-1}^U / s_{ft-1}^U} \right\} \right)^{\frac{1}{\sigma_g^F - 1} - 1}. \quad (A.3.56)
\]

Using equation (A.3.56) together with equations (A.3.17) and (A.3.18), we can construct two moment conditions analogous to those in equation (A.3.29) that can be used to estimate the elasticity of substitution across firms (\(\sigma_g^F\)). Following the same line of reasoning, we can also construct two moment conditions analogous to those in equation (A.3.42) that can be used to estimate the elasticity of substitution across sectors (\(\sigma_g^G\)). We use these alternative representations of the moment conditions as a robustness check for our estimates of the firm and sector elasticities of substitution (\(\sigma_g^F, \sigma_g^G\)) from equations (A.3.29) and (A.3.42). As demand shocks become small (\(\varphi_{igt}^F / \varphi_{igt-1}^F \to 1\) and \(\varphi_{igt}^G / \varphi_{igt-1}^G \to 1\)), or as the number of common varieties becomes large (\(N_{igt,t-1}^F \to \infty\) and \(N_T \to \infty\)) for independently and identically distributed demand shocks, these alternative representations of the moment conditions yield the same estimated elasticities of substitution (\(\sigma_g^F, \sigma_g^G\)). In our empirical results for the U.S. and Chile, we use our baseline specifications in equations (A.3.13) and (A.3.29) for the firm and product elasticities of substitution. We use the robustness specification based on equation (A.3.56) for our sector elasticity of substitution in order to use more of the model’s nesting structure where we have a relatively small number of observations on sectors.

As another robustness check, we use the property of CES that the reverse-weighting estimator can be implemented for any subset of common goods. We now illustrate this property for the firm price index, but it also holds for each of our other tiers of utility. We start by noting that the change in the firm price index for common products (\(\frac{p_{ft}^F}{p_{ft-1}^F}\)) can be written in terms of the change in the firm price index for a subset of common products (\(\frac{p_{ft}^{#}}{p_{ft-1}^{#}}\)) and the change in the expenditure share of this subset in total expenditure on common products (\(\frac{\lambda_{ft}^{U_L}}{\lambda_{ft-1}^{U_L}}\)):

\[
\frac{p_{ft}^F}{p_{ft-1}^F} = \left( \frac{\lambda_{ft}^{U_L}}{\lambda_{ft-1}^{U_L}} \right)^{\frac{1}{\sigma_g^F - 1}} \left[ \frac{\sum_{u \in \Omega_{ft}^{U_L}} \left( \frac{p_{lt}^U / p_{lt-1}^U}{p_{lt-1}^U / p_{lt-1}^U} \right)^{1 - \sigma_g^F}}{\sum_{u \in \Omega_{ft-1}^{U_L}} \left( \frac{p_{lt}^U / p_{lt-1}^U}{p_{lt-1}^U / p_{lt-1}^U} \right)^{1 - \sigma_g^F}} \right]^{\frac{1}{\sigma_g^F - 1}} = \left( \frac{\lambda_{ft}^{U_L}}{\lambda_{ft-1}^{U_L}} \right)^{\frac{1}{\sigma_g^F - 1}} \frac{p_{ft}^{#}}{p_{ft-1}^{#}}, \quad (A.3.57)
\]

where the superscript \(^{#}\) indicates that a variable is defined for this subset of common goods; we denote this subset of common goods by \(\Omega_{ft}^{U_L} \subset \Omega_{ft-1}^{U_L}\); and the shares of expenditure on this subset of common goods in periods \(t - 1\) and \(t\) are:

\[
\lambda_{ft}^{U_L} = \frac{\sum_{u \in \Omega_{ft}^{U_L}} \left( \frac{p_{lt}^U / q_{lt}^U}{q_{lt-1}^U} \right)^{1 - \sigma_g^L}}{\sum_{u \in \Omega_{ft-1}^{U_L}} \left( \frac{p_{lt-1}^U / q_{lt-1}^U}{q_{lt-1}^U} \right)^{1 - \sigma_g^L}}, \quad \lambda_{ft-1}^{U_L} = \frac{\sum_{u \in \Omega_{ft-1}^{U_L}} \left( \frac{p_{lt-1}^U / q_{lt-1}^U}{q_{lt-1}^U} \right)^{1 - \sigma_g^L}}{\sum_{u \in \Omega_{ft-1}^{U_L}} \left( \frac{p_{lt-1}^U / q_{lt-1}^U}{q_{lt-1}^U} \right)^{1 - \sigma_g^L}}. \quad (A.3.58)
\]

Using this property of CES, we obtain the following three equivalent expressions for the change in the firm price index for common products (\(\frac{p_{ft}^F}{p_{ft-1}^F}\)), which are analogous to equations (A.3.1)-(A.3.3) above for the change in the overall firm price index (\(\frac{p_{ft}^F}{p_{ft-1}^F}\)):

\[
\frac{p_{ft}^F}{p_{ft-1}^F} = \left( \frac{\lambda_{ft}^{U_L}}{\lambda_{ft-1}^{U_L}} \right)^{\frac{1}{\sigma_g^F - 1}} \left[ \sum_{u \in \Omega_{ft}^{U_L}} \left( \frac{p_{lt}^U / q_{lt}^U}{p_{lt-1}^U / q_{lt-1}^U} \right)^{1 - \sigma_g^L} \right]^{\frac{1}{\sigma_g^F - 1}}, \quad (A.3.59)
\]
\[
\frac{p_{f_{t+1}}}{p_{f_{t}}} = \left( \frac{\lambda_{U_{t}}^f}{\lambda_{U_{t}}^f} \right)^{1 - \frac{1}{\gamma_{y}}} \left[ \sum_{u \in \Omega_{f_{t+1}}^U} S_{ul}^U \left( \frac{p_{ul}^U / q_{ul}^U}{p_{ul}^U / q_{ul}^U} \right)^{-\left(1 - c_{ul}^U\right)} \right]^{-\frac{1}{\gamma_{y}}},
\]  \hspace{1cm} (A.3.60)

\[
\frac{p_{f_{t+1}}}{p_{f_{t}}} = \left( \frac{\lambda_{U_{t}}^f}{\lambda_{U_{t}}^f} \right)^{1 - \frac{1}{\gamma_{y}}} M_{f_{t}}^{U_{t}} \left[ \frac{p_{ul}^U / p_{ul}^{U_{t}}}{p_{ul}^U / p_{ul}^{U_{t}}} \right] \left( M_{f_{t}}^{U_{t}} \left[ \frac{S_{ul}^U / S_{ul}^{U_{t}}}{S_{ul}^U / S_{ul}^{U_{t}}} \right] \right)^{1 - \frac{1}{\gamma_{y}}}, \hspace{1cm} (A.3.61)
\]

where \(S_{ul}^{U_{t-1}}\) is the share of an individual product in total expenditure on this subset of common goods:

\[
S_{ul}^{U_{t-1}} = \frac{(p_{ul}^U / q_{ul}^U)^{1 - c_{ul}^U}}{\sum_{t \in \Omega_{f_{t+1}}^U} (p_{ul}^U / q_{ul}^U)^{1 - c_{ul}^U}}, \hspace{1cm} u \in \Omega_{f_{t+1}}^U. \hspace{1cm} (A.3.62)
\]

\(M_{f_{t}}^{U_{t}} [ \cdot ]\) is the geometric mean across this subset of common goods such that:

\[
M_{f_{t}}^{U_{t}} \left[ p_{ul}^U / q_{ul}^U \right] = \left( \prod_{u \in \Omega_{f_{t+1}}^U} p_{ul}^U \right)^{1 / N_{f_{t+1}}^U} = 1, \hspace{1cm} (A.3.63)
\]

where \(N_{f_{t+1}}^U = \left| \Omega_{f_{t+1}}^U \right|\) is the number of elements in this subset of common goods; and we now choose units in which to measure product demand \((q_{ul}^U)\) such that its geometric mean across this subset of common goods is equal to one:

\[
M_{f_{t}}^{U_{t}} \left[ q_{ul}^U \right] = \left( \prod_{u \in \Omega_{f_{t+1}}^U} q_{ul}^U \right)^{1 / N_{f_{t+1}}^U} = 1. \hspace{1cm} (A.3.64)
\]

Using the three equivalent expressions for the change in each firm’s price index in equations (A.3.59)-(A.3.61), and re-arranging terms, we obtain the following two equalities:

\[
\begin{align*}
\Theta_{f_{t+1}}^{U_{t-1}} & = \left[ \sum_{u \in \Omega_{f_{t+1}}^U} S_{ul}^{U_{t-1}} \left( \frac{p_{ul}^U}{p_{ul}^{U_{t}} - 1} \right)^{1 - c_{ul}^U} \right]^{1 - \frac{1}{\gamma_{y}}} = M_{f_{t}}^{U_{t}} \left[ \frac{p_{ul}^U / p_{ul}^{U_{t}}}{p_{ul}^U / p_{ul}^{U_{t}}} \right] \left( M_{f_{t}}^{U_{t}} \left[ \frac{S_{ul}^U / S_{ul}^{U_{t}}}{S_{ul}^U / S_{ul}^{U_{t}}} \right] \right)^{1 - \frac{1}{\gamma_{y}}}, \hspace{1cm} (A.3.65)
\end{align*}
\]

\[
\begin{align*}
\left( \Theta_{f_{t+1}}^{U_{t-1}} \right)^{-1} & = \left[ \sum_{u \in \Omega_{f_{t+1}}^U} S_{ul}^{U_{t-1}} \left( \frac{p_{ul}^U}{p_{ul}^{U_{t}} - 1} \right)^{1 - c_{ul}^U} \right]^{-\frac{1}{\gamma_{y}}} = M_{f_{t}}^{U_{t}} \left[ \frac{p_{ul}^U / p_{ul}^{U_{t}}}{p_{ul}^U / p_{ul}^{U_{t}}} \right] \left( M_{f_{t}}^{U_{t}} \left[ \frac{S_{ul}^U / S_{ul}^{U_{t}}}{S_{ul}^U / S_{ul}^{U_{t}}} \right] \right)^{-\frac{1}{\gamma_{y}}}, \hspace{1cm} (A.3.66)
\end{align*}
\]

where the terms in the share of expenditure on this subset of common products \((\lambda_{U_{t}}^f / \lambda_{U_{t}}^f)^{1 / (c_{ul}^U - 1)}\) have cancelled; \(\Theta_{f_{t+1}}^{U_{t-1}}\) is a forward aggregate demand shifter and \(\Theta_{f_{t+1}}^{U_{t-1}}\) is a backward aggregate demand shifter such that:

\[
\Theta_{f_{t+1}}^{U_{t+1}} \equiv \left[ \sum_{u \in \Omega_{f_{t+1}}^U} S_{ul}^{U_{t-1}} \left( \frac{p_{ul}^U / p_{ul}^{U_{t}}}{p_{ul}^U / p_{ul}^{U_{t}}} \right)^{1 - c_{ul}^U} \right]^{-\frac{1}{\gamma_{y}}}, \hspace{1cm} (A.3.67)
\]

\[
\Theta_{f_{t+1}}^{U_{t-1}} = \left[ \sum_{u \in \Omega_{f_{t+1}}^U} S_{ul}^{U_{t-1}} \left( \frac{p_{ul}^U / p_{ul}^{U_{t}}}{p_{ul}^U / p_{ul}^{U_{t}}} \right)^{1 - c_{ul}^U} \right]^{-\frac{1}{\gamma_{y}}}. \hspace{1cm} (A.3.68)
\]
Using the identifying assumption that the demand shocks cancel out across this subset of common products
\((\Theta^{1\#}_{t1, t-1} = (\Theta^{t1\#}_{t1, t-1})^{-1} = 1)\), equations (A.3.67) and (A.3.68) can be used to construct moment conditions to estimate the elasticity of substitution across products \((\sigma^U_g)\) that are analogous to those in equation (A.3.13) above. In estimating the elasticities of substitution for the U.S. and Chile, we focus on the subset of common goods for each tier of utility \(K\) that have relative changes in prices \((P^K / P^{K}_{1-t})\) and expenditures \((X^K / X^{K}_{1-t})\) in between the 10th and 90th percentiles, which enables us to abstract from implausibly large annual changes in prices and expenditures for outlying observations. Given these estimated elasticities of substitution \((\sigma^U_g, \sigma^F_g, \sigma^G)\), we solve for the demand shifters \((\phi^U_{u}, \phi^F_{f}, \phi^G_{jg})\) that rationalize the observed data on prices \((P^U_{ut})\) and expenditures \((X^U_{ut})\) for all observations.

### A.4 Data Description

In this section of the web appendix, we report further details on the data sources and definitions for the U.S. trade transactions data and Chilean trade transactions data used in the paper.

#### A.4.1 U.S. Data

The U.S. trade transactions data comes from the U.S. Census Bureau’s Longitudinal Firm Trade Transactions Database (LFTTD). This database covers the universe of U.S.-based firms that import merchandise from abroad. For each import shipment, we observe the freight value of the shipment in U.S. dollars, the quantity shipped, the date of the transaction, the product classification (according to 10-digit Harmonized System (HS) codes), and the Manufacturing ID (MID). The MID is a field that importing firms must record in CBP Form 7501 in order to complete the importation of goods into the United States.

We use the MID to identify the manufacturer of the merchandise. The first two characters of the MID are the two-digit ISO country code for the country of origin. The next three characters are the start of the first word of the exporter’s name. The next three characters are the start of the second word of that name. The next four characters are the start of the largest number that appears in the street address of the exporter. The last three characters are the start of the exporter’s city.

Kamal, Krizan and Monarch (2015) documents the characteristics of the MID and its ability to identify a foreign supplier. The authors show that simple cleaning procedures, such as removing the city portion of the MID or removing the address-number portion of the MID, result in a close match between the number of exporting firms to the U.S. from each exporting country reported in the LFTTD and that reported in exporting country data.

Guided by these results, we define foreign exporting firms using the MID, after having removed both the address-number and the city, and the NAICS 4-digit code. This procedure enables us to merge together multi-plant firms that operate in different cities. After implementing this procedure, we compared the number of firms per country exporting to the U.S. in the LFTTD and foreign country sources and found that they matched closely. In addition to removing the address from the MID, we also implement the following additional cleaning procedures:
1. Standardize the units in which quantities are reported (e.g., we convert dozens to counts and grams to kilograms).

2. Drop an observation if the unit of quantity does not exist.

3. Drop observations that are indicated to have a high likelihood of input error (as indicated by a "blooper" variable in the data).

4. Drop an observation if the MID is missing.

5. Drop an observation if the ISO code (the first two digits of the MID) is invalid.

6. Drop an observation if the MID does not contain the firm-name portion.

7. Drop an observation if the quantity or value is invalid (negative or missing).

8. If the exporter is from Canada, the first two letters in the MID denotes the Canadian province rather than the ISO code of Canada. We therefore collapse provinces into one Canada.

9. The ISO codes in the MID often separate China and Hong Kong, which we collapse into China.

10. Our transaction data includes imports from U.S. territories and also imports from domestic origin returned to the United States with no change in condition or after having been processed and/or assembled in other countries. We drop these observations, so that we only consider transactions with a foreign country of origin.

A.4.2 Chilean Data

The Chilean trade transactions data come from Datamyne and take a similar form as our U.S. trade transactions data. For each import customs shipment, we observe the cost-inclusive-of-freight value of the shipment in U.S. dollars (converted using market exchange rates), the quantity shipped, the date of the transaction, the product classification (according to 8-digit Harmonized System (HS) codes), the country of origin, and the brand of the exporter (e.g., Nestlé, Toyota).

Using this information on import shipments, we construct a dataset for importer \( j \) (Chile) with many exporters \( i \) (countries of origin), sectors \( g \) (2-digit HS codes), firms \( f \) (foreign brands within exporter within sector), and products \( u \) (8-digit HS codes within foreign brands within sectors) and time \( t \) (year). We drop the small number of HS8 codes that do not use consistent units over time (e.g. we drop any HS8 code that switches from counts to kilograms). We also drop any observations for which countries of origin or brands are missing as well as those where the brand is a major trading company.\(^1\) After several additional cleaning rules, which will be outlined in the next section, we collapse the import shipments data to the annual level by exporting firm and product, weighting by trade value, which yields a total of 6.5 million observations on Chilean imports by exporter-firm-product-year spanning the years 2007-2014.

\(^1\)These were taken from the Forbes list of the top 10 trading companies.
A.4.3 Data Cleaning Methodology for Chilean Data

In this section, we explain the method used to clean and cluster the firm names in the Chilean import data.

A.4.3.1 Initial Cleaning of Raw Firm Names

We begin by implementing the following basic cleaning procedures to deal with obvious and easily fixable problems with the firm names.

1. Drop trading company names such as “MITSUBISHI CORPORATION”, “MITSUBISHI CORP”, and “SUM-ITOMO CORP”.
2. Trim company names to have a maximum string length of 50 (this impacted two firm names).
3. Remove substrings such as “-F”, “- F”, “S.A.”.
4. Remove most punctuations and symbols. We remove all of the following: ;,:()[{}!%?/\@^*
5. Drop firm names that consist of only one alphabetical letter (e.g. if the brand name is “A”).
6. Add a space in front of common words. We implement this, because we observe many conjoined words (e.g. APPLEINCORPORATED).
7. Remove extra spaces between words (when there is more than one space between words) and remove spaces that come before or after the firm name (e.g. “APPLE INCORPORATED” becomes “APPLE INCORPORATED”).
8. Delete companies that are identified only by a Chinese city name (e.g. firm name is simply “BEIJING”).
9. After applying these steps, remove firm names that are blank.

A.4.3.2 Standardizing Firm Names

We then use stnd_compname, a user-written Stata package by Wasi and Flaaen primarily to:

1. Remove entity names (e.g. LLC, LTD, INC)
2. Shorten commonly used words (e.g. ELECTRONICS, TECHNOLOGY) that have less distinguishing power, so that they will have less weighting during the string-similarity clustering.

The stnd_compname package comes with 43 standardizations for approximately 104 commonly used words (such as ENTERPRISE, INTERNATIONAL, MANAGEMENT, etc). We add approximately 100 standardizations and 180 words to this list for a total of 150 standardizations and around 300 words based on which words were the most common in the data. In addition to standardizing words, we also implement two more cleaning steps to complement the standardization:

1. Search through and remove a word if the first letter of the word is a numeral and the word is not the first word of the firm name. (MAZDA 4X 7TR turns into MAZDA)
2. If there is numeral within a word that is not the first word in a name, we remove the numeral and the rest of the letters following the numeral in the word. (FUJI F342FDIF turns into FUJI F)

A.4.3.3 Clustering

We then run string-similarity clustering (using strgroup, a user-written Stata package by Julian Reif) on the standardized firm names using a number of different thresholds and groups. These thresholds determine two strings’ edit distance below which the two strings (i.e. firm names) will be grouped together. Varying this threshold is useful, because we observe that firm names are more likely to refer to the same firm if they share the same HS category. For example, we would be more comfortable assuming that “Sony Corp” and “Pony Corp” refer to the same company if we were only looking at makers of DVD players than doing cross-sector comparisons (because such cross-sector comparisons could involve assuming that an exporter of DVDs is also an exporter of farm animals). We take advantage of this by implementing clustering multiple times within multiple HS levels (2, 4, 6 and 8) and choosing stricter clustering thresholds for broader HS levels (i.e. as we cluster within more disaggregated HS-levels, the criterion for grouping firm names are made less strict). Specifically, we set our thresholds at 15 percent, 20 percent, 22 percent, and 30 percent for clustering within 2-digit HS codes, 4-digit HS codes, 6-digit HS codes, and 8-digit HS codes, respectively. After creating 4 different firm identifiers for the various HS levels (HS2, HS4, HS6, and HS8), the groupings are then merged together. If firm name A is matched with firm name B, and firm name B is matched with firm name C, then firm name A is matched with firm name C, and so on.

In parallel with the string-similarity clustering on the standardized firm names, we also implement the string similarity clustering on the firm names prior to standardization. We do this in case the standardization was ineffective (e.g. we missed certain words to be standardized). We run this clustering on a much stricter threshold than in the earlier step, so that we remain conservative about grouping firm names together. If the clustering results are too large (i.e. the threshold is not strict enough), we restrict the size of a cluster to 5 unique firm names (so that a firm name can be spelled in up to 5 different ways while still be identified as the same firm).

After clustering on the two sets of firm names (the firm names prior to standardization and those after standardization) we merge the clusters together. If firm A is matched with firm B in the first step and firm B is matched with firm C in the second step, then these groupings are merged, so that firm A is matched with C as well, implying that firms A, B, and C are all allocated to the same group.

A.4.3.4 Additional Cleaning Steps

After standardizing and clustering, we apply additional cleaning rules:

1. Now that standardization and clustering is complete, we drop the remaining observations with trading companies, blank firm names, and firm names that are only identified by a single alphabetical letter.

2. We observe many firm names in the data of the form “A & W” or “T & W” where the firm names consist of two letters with an “&” in between. The clustering method often clusters these firm names together.
(depending on the HS level) even if only one of these letters are the same (e.g. “A & W” and “T & W”), because the difference between the two firm names are 1/5 or 20 percent, which is within the threshold in many cases. To address this, we apply a rule such that these firm names are separated into different groups unless there is an exact match.

3. We again restrict the size of a cluster to 5 unique firm names. If a cluster is larger than 5 unique firm names, we cluster again on an ever-stricter threshold until the size of the cluster is five or less.

4. We sometimes encounter observations where the entire firm name is contained exactly at the start of another (e.g. "SONY" and "SONY ELECTRONICS" or "HEWLETT PACKARD" and "HEWLETT PACKARD ENTERPRISE"). Even after standardizing common words, these firm names often fail to be clustered together because their edit distances are too large. We combat this by creating a rule such that if one firm name appears at the beginning of another, the two firm names are grouped together.

A.4.3.5 Validation

After implementing the above steps, we then checked how well our procedure worked by manually checking the results of this algorithm for the 1,249 raw firm names in the Japanese steel sector (which we had not looked at when developing the procedure). We manually checked the accuracy using two steps. First, we sorted the firm list alphabetically and counted the number of firm names that should have been grouped together (based on our manual inspection) but were not grouped together by our clustering algorithm. Second, we sorted the firm names by our groups and counted the number of firm names that should not have been grouped together (based on our manual inspection) but were grouped together by our clustering algorithm. Summing these type I and type II errors, we found that our cleaning algorithm and manual checking grouped firms in the same way for 99.9 percent of observations. As a final check on the sensitivity of our results to this cleaning algorithm, we replicated our main results of the Chilean import transactions data using the firm names prior to these cleaning steps. Again we find that most of the variation in revealed comparative advantage (RCA) across countries and sectors is explained by variety and demand/quality. Therefore, while our clustering algorithm improves the allocation of import transactions to firms, our main qualitative and quantitative conclusions hold regardless of whether or not we use this algorithm.

A.5 U.S. Empirical Results

In this section of the web appendix, we report additional empirical results using our U.S. data for Section 5 of the paper.

A.5.1 Elasticities of Substitution

In Figure A.5.1, we plot our estimated product, firm and sector elasticities of substitution \( \hat{\sigma}^{U}_g, \hat{\sigma}^{F}_g, \hat{\sigma}^{G}_g \), sorted based on the ranking of the estimated firm elasticity of substitution \( \hat{\sigma}^{F}_g \). We also show 95% confidence intervals for the estimated product and firm elasticities of substitution \( \hat{\sigma}^{U}_g, \hat{\sigma}^{F}_g \) based on bootstrapped standard errors. As can be seen in the figure, we find a natural ordering where \( \hat{\sigma}^{U}_g > \hat{\sigma}^{F}_g > \hat{\sigma}^{G}_g \). We also find that the
confidence intervals are narrow enough such that the product elasticity is significantly larger than the firm elasticity ($\hat{\sigma}^{UI}_{g} > \hat{\sigma}^{F}_{g}$) at the 5 percent level of significance for all sectors, and the firm elasticity is significantly larger than the sector elasticity ($\hat{\sigma}^{F}_{g} > \hat{\sigma}^{G}$) at this significance level for all sectors.

![Estimated Elasticities of Substitution](image)

**Figure A.5.1**: Estimated Elasticities of Substitution, Within Firms ($\hat{\sigma}^{UI}_{g}$), Across Firms ($\hat{\sigma}^{F}_{g}$) and Across Sectors ($\hat{\sigma}^{G}$), sorted based on the ranking of $\hat{\sigma}^{F}_{g}$ (U.S. Data)

### A.5.2 Exporter Price Indexes Across Sectors and Countries

No further results required.

### A.5.3 Trade Patterns

As discussed in Sections 5.1 and 5.3 of the paper, we undertake a robustness check, in which we carry out a grid search over the range of plausible values for elasticities of substitution across firms and products. In particular, we consider values of $\sigma^{F}_{g}$ from 2 to 8 (in 0.5 increments) and values of $\sigma^{UI}_{g}$ from $(\sigma^{F}_{g} + 0.5)$ to 20 in 0.5 increments, while holding $\sigma^{G}$ constant at our estimated value, which respects our estimated ranking that $\sigma^{UI}_{g} > \sigma^{F}_{g} > \sigma^{G}$.

We begin by showing that the percentage contributions from firm variety and firm dispersion are invariant across this parameter grid, because the elasticities of substitution cancel from these expressions. From equation (A.2.42) in Section A.2.10.1 of this web appendix, the overall contribution from both firm and product variety to the level of log RCA is,
\[
\ln \left( \text{RCA}^A_{\text{jit}, t} \right) = - \left\{ \begin{array}{l}
\Delta \ln \lambda^A_{\text{jit}, t} - \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^A_{j|t-1}} \Delta \ln \lambda^A_{h|t, t} \\
- \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^A_{j|t-1}} \lambda^A_{h|t, t} \cdot \sum_{h \in \Omega^A_{j|t-1}} \Delta \ln \lambda^A_{h|t, t} \\
+ \frac{c^A_{j|t-1}}{c^A_{j|t-1}} \sum_{h \in \Omega^A_{j|t-1}} \lambda^A_{h|t, t} \cdot \sum_{h \in \Omega^A_{j|t-1}} \Delta \ln \lambda^A_{h|t, t} \\
\end{array} \right\},
\]

where the component of this contribution that captures firm variety is,

\[
\ln \left( \text{RCA}^F_{\text{jit}, t} \right) = - \left\{ \begin{array}{l}
\Delta \ln \lambda^F_{\text{jit}, t} - \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^F_{j|t-1}} \Delta \ln \lambda^F_{h|t, t} \\
- \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^F_{j|t-1}} \lambda^F_{h|t, t} \cdot \sum_{h \in \Omega^F_{j|t-1}} \Delta \ln \lambda^F_{h|t, t} \\
+ \frac{c^F_{j|t-1}}{c^F_{j|t-1}} \sum_{h \in \Omega^F_{j|t-1}} \lambda^F_{h|t, t} \cdot \sum_{h \in \Omega^F_{j|t-1}} \Delta \ln \lambda^F_{h|t, t} \\
\end{array} \right\},
\]

which depends solely on observed moments in the data and is invariant to the assumed elasticities of substitution for finite values of these elasticities \((\sigma^U_G < \infty\) and \(\sigma^F_G < \infty\)). Taking differences over time in equation (A.5.2), this invariance result also holds for changes in log RCA.

Similarly, from equation (A.2.41) in Section A.2.10.1 of this web appendix, the overall contribution from the dispersion of demand-adjusted prices across common products and firms for the level of log RCA is,

\[
\ln \left( \text{RCA}^S_{\text{jit}, t} \right) = - \left\{ \begin{array}{l}
\Delta \ln S^S_{\text{jit}, t} - \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^S_{j|t-1}} \Delta \ln S^S_{h|t, t} \\
- \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^S_{j|t-1}} S^S_{h|t, t} \cdot \sum_{h \in \Omega^S_{j|t-1}} \Delta \ln S^S_{h|t, t} \\
+ \frac{c^S_{j|t-1}}{c^S_{j|t-1}} \sum_{h \in \Omega^S_{j|t-1}} S^S_{h|t, t} \cdot \sum_{h \in \Omega^S_{j|t-1}} \Delta \ln S^S_{h|t, t} \\
\end{array} \right\},
\]

where the component of this contribution that captures firm dispersion is,

\[
\ln \left( \text{RCA}^S_{\text{ jit}, t} \right) = - \left\{ \begin{array}{l}
\Delta \ln S^S_{\text{ jit}, t} - \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^S_{j|t-1}} \Delta \ln S^S_{h|t, t} \\
- \frac{1}{N_{j|t-1}} \sum_{h \in \Omega^S_{j|t-1}} S^S_{h|t, t} \cdot \sum_{h \in \Omega^S_{j|t-1}} \Delta \ln S^S_{h|t, t} \\
+ \frac{c^S_{j|t-1}}{c^S_{j|t-1}} \sum_{h \in \Omega^S_{j|t-1}} S^S_{h|t, t} \cdot \sum_{h \in \Omega^S_{j|t-1}} \Delta \ln S^S_{h|t, t} \\
\end{array} \right\},
\]

which depends solely on observed moments in the data and is invariant to the assumed elasticities of substitution for finite values of these elasticities \((\sigma^U_G < \infty\) and \(\sigma^F_G < \infty\)). Taking differences over time in equation (A.5.4), this invariance result also again holds for changes in log RCA.

In Figure A.5.2, we show histograms across the parameter grid for the contribution from each of the remaining terms from our decomposition of the level of RCA in equation (31) in the paper. The contributions from product prices, product variety and product dispersion in the final three panels sum to the contribution from firm prices in the first panel. Additionally, the firm price and firm demand contributions in the first two panels plus the unreported contributions from firm variety and firm dispersion sum to one. In Figure A.5.3, we display analogous results for our decomposition of changes in RCA over time, where the five panels of the figure have the same relationship with one another as in Figure A.5.2.

In both figures, a higher value for \(\sigma^F_G\) raises the contribution from average prices and reduces the contribution from average demand/quality. Nonetheless, across the entire grid of parameter values, average prices account for less than 30 percent of the level of the RCA and less than 10 percent of the changes in RCA. In contrast, for all parameter values on the grid, average demand’s contribution to the level of RCA is around as
large as that from average prices (from less than 5 percent to over 25 percent in Figure A.5.2). Furthermore, its contribution to changes in RCA is substantially larger than that from average prices (from just over 35 percent to just under 60 percent in Figure A.5.3).

In summary, our findings that most of the variation in patterns of RCA is explained by factors other than average prices is robust to the consideration of alternative elasticities of substitution. In particular, the contributions from firm variety and firm dispersion are invariant to these elasticities of substitution. Furthermore, across the range of plausible values for these elasticities of substitution, the contribution from average demand remains large relative to that from average prices.

Figure A.5.2: Contributions to the Level of U.S. RCA in 2011 Across the Parameter Grid for the Firm and Product Elasticities of Substitution
A.5.3 Contributions to the Change in U.S. RCA in from 1998-2011 Across the Parameter Grid for the Firm and Product Elasticities of Substitution

A.5.4 Additional Theoretical Restrictions

In Section 5.4 of the paper, we compare the observed data for firm sales and our model solutions for the firm price index and firm demand/quality ($\ln V_{ft}^F \in \{\ln X_{ft}^F, \ln P_{ft}^F, \ln q_{ft}^F\}$) with their theoretical predictions under the assumptions of an untruncated Pareto distribution or a log normal distribution. In this section of the web appendix, we derive these theoretical predictions, as summarized in equations (32) and (33) in the paper.

**Empirical Distributions** In particular, we use the QQ estimator of Kratz and Resnick (1996), as introduced into the international trade literature by Head, Mayer and Thoenig (2016). We start with the empirical distributions. Ordering firms by the value of a given variable $V_{ft}^F$ for $f \in \{1, \ldots, N_{jigt}^F\}$ for a given exporter $i$ to importer $j$ in sector $g$ at time $t$, we observe the empirical quantiles:

$$V_{ft} = \ln \left( V_{ft}^F \right). \quad (A.5.5)$$

We can use these empirical quantiles to estimate the empirical cumulative distribution function:

$$\hat{F}_{jigt} \left( V_{ft}^F \right) = \frac{f - b}{N_{jigt}^F + 1 - 2b}, \quad b = 0.3, \quad (A.5.6)$$

where the plot position of $b = 0.3$ can be shown to approximate the median rank of the distribution (see Benard and Boslevenbach 1953). We next turn to the theoretical distributions, first under the assumption of an untruncated Pareto distribution, and next under the assumption of a log normal distribution.
Untruncated Pareto Distribution Under the assumption that the variable $V_{ft}^E$ has an untruncated Pareto distribution, its cumulative distribution function is given by:

$$F_{ijgt}(V_{ft}^E) = 1 - \left( \frac{V_{ijgt}^E}{V_{ft}^E} \right)^{a_g^V}, \quad (A.5.7)$$

where $F_{ijgt} (\cdot)$ is the cumulative distribution function; $V_{ijgt}^E$ is the lower limit of the support of the distribution for variable $V_{ft}^E$ for exporter $i$, importer $j$, sector $g$ and time $t$; and $a_g^V$ is the Pareto shape parameter for variable $V_{ft}^E$ for sector $g$.

Inverting this cumulative distribution function, and taking logarithms, we obtain the following predicted theoretical quantile for each variable:

$$\ln (V_{ft}^E) = \ln V_{ijgt}^E - \frac{1}{a_g^V} \ln \left[ 1 - F_{ijgt}(V_{ft}^E) \right], \quad (A.5.8)$$

which corresponds to equation (32) in the paper.

We estimate equation (A.5.8) by OLS using the empirical quantile from equation (A.5.5) for $\ln (V_{ft}^E)$ on the left-hand side and the empirical estimate of the cumulative distribution function from equation (A.5.6) for $F_{ijgt}(V_{ft}^E)$ on the right-hand side. We estimate this regression for each sector across foreign firms (allowing the slope coefficient $a_g^V$ to vary across sectors) and including fixed effects for each exporter-sector-year combination (allowing the intercept $\ln V_{ijgt}^E$ to vary across exporters, sectors and time). The fitted values from this regression correspond to the predicted theoretical quantiles, which we compare to the empirical quantiles observed in the data. Under the null hypothesis of a Pareto distribution, there should be a linear relationship between the theoretical and empirical quantiles that coincides with the 45-degree line.

To assess this theoretical prediction, we estimate equation (32) in the paper for two separate subsamples: firms with values below the median for each exporter-sector-year cell and firms with values above the median for each exporter-sector-year cell. Under the null hypothesis of a Pareto distribution, the estimated slope coefficient $1/a_g^V$ should be the same for firms below and above the median. In the bottom three panels of Figure A.5.4, we display the estimated slope coefficients $1/a_g^V$ for each 4-digit NAICS industry for the log firm price index ($\ln P_{ft}^F$ to the left), log firm exports ($\ln X_{ft}^E$ in the middle), and log firm demand ($\ln q_{ft}^E$ to the right). In each panel, we sort industries by the estimated slope coefficient for the full sample for that variable (shown by the black straight line). The red and blue numeric industry codes show the estimates for the subsamples of firms below and above the median respectively. For all three variables, we strongly reject the null hypothesis of a Pareto distribution, with substantial differences in the estimated coefficients below and above the median, which are significant at conventional levels.

Log Normal Distribution In contrast, under the assumption that the variable $V_{ft}^E$ has a log normal distribution, its cumulative distribution function is given by:

$$\ln (V_{ft}^E) \sim \mathcal{N}(\kappa_{ijgt}^V, (\chi_g^V)^2), \quad (A.5.9)$$
Figure A.5.4: Estimated Coefficients from Regressions of the Empirical Quantiles on the Theoretical Quantiles Implied by a Pareto or Log Normal distribution (U.S. data)

where $\kappa^V_{jigt}$ is the mean for $\ln V_{ft}^E$ for exporter $i$ in importer $j$ and sector $g$ at time $t$ and $\chi^V_g$ is the standard deviation for $\ln V_{ft}^E$ for sector $g$. It follows that the standardized value of the log of each variable is drawn from a standard normal distribution:

$$F_{jigt} (V_{ft}^E) = \Phi \left( \frac{\ln (V_{ft}^E) - \kappa^V_{jigt}}{\chi^V_g} \right), \tag{A.5.10}$$

where $\Phi (\cdot)$ is the standard normal cumulative distribution function. Inverting this cumulative distribution function, we obtain the following predictions for the theoretical quantiles of each variable:

$$\frac{\ln (V_{ft}^E) - \kappa^V_{jigt}}{\chi^V_g} = \Phi^{-1} \left( F_{jigt} \left( V_{ft}^E \right) \right), \tag{A.5.11}$$

which can be re-expressed as:

$$\ln \left( V_{ft}^E \right) = \kappa^V_{jigt} + \chi^V_g \Phi^{-1} \left( F_{jigt} \left( V_{ft}^E \right) \right), \tag{A.5.12}$$

which corresponds to equation (33) in the paper.

Again we estimate equation (A.5.12) by OLS using the empirical quantile from equation (A.5.5) for $\ln \left( V_{ft}^E \right)$ on the left-hand side and the empirical estimate of the cumulative distribution function from equation (A.5.6).
for $F_{ijgt} (V^F_{jt})$ on the right-hand side. We estimate this regression for each sector across foreign firms (allowing the slope coefficient $\chi^V_S$ to vary across sectors) and including fixed effects for each exporter-sector-year combination (allowing the intercept $\kappa^V_{ijgt}$ to vary across exporters, sectors and time). In the top three panels of Figure A.5.4, we display the estimated slope coefficients $\chi^V_S$ for each 4-digit NAICS industry for the log firm price index ($\ln P^F_{jt}$ to the left), log firm exports ($\ln X^F_{jt}$ in the middle), and log firm demand ($\ln q^F_{jt}$ to the right), using the same coloring as for the bottom three panels discussed above. As apparent from the figure, we find that the log normal distributional assumption provides a closer approximation to the data than the Pareto distributional assumption. Consistent with Bas, Mayer and Thoenig (2017), we find smaller departures from the predicted linear relationship between the theoretical and empirical quantiles for a log normal distribution than for a Pareto distribution. Nevertheless, we reject the null hypothesis of a log normal distribution at conventional significance levels for all three variables for the majority of industries, with substantial differences in estimated coefficients above and below the median for a number of industries. Instead of imposing such supply-side distributional assumptions, our demand-side approach uses the observed empirical distributions of prices and expenditure shares, and the resulting implied distribution of demand/quality under our assumption of CES demand.

### A.5.5 Additional Reduced-Form Evidence

In Figures A.5.5-A.5.8 below, we show that our U.S. trade transactions data exhibit have the same reduced-form properties as found in existing studies in the empirical trade literature (see for example Bernard, Jensen and Schott 2009 and Bernard, Jensen, Redding and Schott 2009 for the U.S.; Mayer, Melitz and Ottaviano 2014 for France; and Manova and Zhang 2012 for China).

First, we find a high concentration of trade across countries and a dramatic increase in Chinese import penetration over time. As shown in Figure A.5.5, the top 20 import source countries account for around 80 percent of U.S. imports; China’s import share more than doubles from 7 to 18 percent from 1997-2011; in contrast, Japan’s import share more than halves from 14 to 6 percent over this period.

Second, we find high rates of product and firm turnover and evidence of selection conditional on product and firm survival. In Figure A.5.6, we display the fraction of firm-product observations and import value by tenure (measured in years) for 2011, where recall that firms here correspond to foreign exporting firms. Around 50 percent of the firm-product observations in 2011 have been present for two years or less, but the less than 5 percent of these observations that have survived for at least fifteen years account for over 20 percent of import value.

Third, we find that international trade is dominated by multi-product firms. In Figure A.5.7, we display the fraction of firm observations and import value in 2011 accounted for by firms exporting different numbers of products. Although less than 40 percent of exporting firms are multi-product, they account for more than 90 percent of import value.

Fourth, we find that the extensive margins of firm and product exporting account for most of the cross-section variation in aggregate trade. In Figure A.5.8, we display the log of the total value of U.S. imports from each foreign country, the log number of firm-product observations with positive trade for that country, and the
log of average imports per firm-product observation with positive trade from that country. We display these three variables against the rank of countries in U.S. total import value, with the largest country assigned a rank of one (China). By construction, total import value falls as we consider countries with higher and higher ranks. Substantively, most of this decline in total imports is accounted for by the extensive margin of the number of firm-product observations with positive trade, whereas the intensive margin of average imports per firm-product observation with positive trade remains relatively flat.

Therefore, across these and a range of other empirical moments, our data are representative of existing empirical findings using international trade transactions data.
Figure A.5.5: Country Shares of U.S. Imports over Time

Figure A.5.6: Distribution Firm-Product Observations and Import Value by Tenure 2011 (U.S. data)

Note: Data are for 2011. Tenure is the number of years a firm-product observation has existed since 1997.
Figure A.5.7: Distribution of Firm Observations Across Number of Products 2011 (U.S. data)

Figure A.5.8: Extensive and Intensive Margins of Firm-Product Imports Across Countries 2011 (U.S. data)
A.6 Chilean Empirical Results

In this section of the web appendix, we replicate the empirical results from Section 5 of the paper, but using our Chilean data instead of our U.S. data. In Section A.6.1, we report our estimates of the elasticities of substitution \((\sigma^U_s, \sigma^F_s, \sigma^G_s)\), which we use to invert the model and recover the values of product, firm and sector demand/quality \((q^U_{it}, q^F_{ft}, q^G_{fgt})\). In Section A.6.2, we use these estimates to compute the exporter price indexes that determine the cost of sourcing goods across countries and sectors. In Section A.6.3, we report our main results for the determinants of comparative advantage, aggregate trade and aggregate prices. In Section A.6.4, we compare the results of our framework with special cases that impose additional theoretical restrictions. Finally, in Section A.6.5, we confirm that our Chilean data exhibit the same reduced-form properties as our U.S. data and as found in other empirical studies using international trade transactions data.

A.6.1 Elasticities of Substitution

We begin by showing that we find a similar pattern of estimated elasticities of substitution using the Chilean data as using the U.S. data in Section 5.1 of the paper. In Table A.1, we summarize our estimates of the elasticities of substitution \((\sigma^U_s, \sigma^F_s, \sigma^G_s)\) using the Chilean data. We report quantiles of the distributions of the estimated product and firm elasticities \((\sigma^U_s, \sigma^F_s)\) across sectors, as well as the single estimated elasticity of substitution across sectors \((\sigma^G_s)\). As for the U.S., we find that the estimated product and firm elasticities are statistically significantly larger than one, and always below eleven. We obtain a median estimated elasticity across products \((\sigma^U_s)\) of 5.0, a median elasticity across firms \((\sigma^F_s)\) of 2.7 and an elasticity across sectors \((\sigma^G_s)\) of 1.69, which compare closely with our U.S. estimates.

Although we do not impose this restriction on the estimation, we again find a natural ordering, in which varieties are more substitutable within firms than across firms, and firms are more substitutable within industries than across industries: \(\hat{\sigma}^U_s > \hat{\sigma}^F_s > \hat{\sigma}^G_s\). We find that the product elasticity is significantly larger than the firm elasticity at the 5 percent level of significance for 98 percent of sectors, and the firm elasticity is significantly larger than the sector elasticity at this significance level for 88 percent of sectors. Therefore, the Chilean data also rejects the special cases in which consumers only care about firm varieties \((\sigma^U_s = \sigma^F_s = \sigma^G_s)\), in which varieties are perfectly substitutable within sectors \((\sigma^U_s = \sigma^F_s = \infty)\), and in which products are equally differentiated within and across firms for a given sector \((\sigma^U_s = \sigma^F_s)\). Instead, we find evidence of both firm differentiation within sectors and product differentiation within firms, as for the U.S. in the paper.

Our estimated elasticities of substitution are again broadly consistent with those of other studies that have used similar data but different methodologies and/or nesting structures. Our estimates of the product and firm elasticities \((\sigma^F_s\) and \(\sigma^U_s)\) are only slightly smaller than those estimated by Hottman et al. (2016) using different data (U.S. barcodes versus internationally-traded HS products) and a different estimation methodology based on Feenstra (1994). As a robustness check, if we apply this alternative methodology to our data, we also obtain quite similar estimates, with median elasticities of 4.2 at the product level and 1.8 at the firm level, which are close to the 5.0 and 2.7 obtained here. Thus, our estimated elasticities do not differ substantially

\(^2\)Our median estimates for the elasticities of substitution within and across firms of 5.0 and 2.7 respectively compare with those of 6.9 and 3.9 respectively in Hottman et al. (2016).
from those obtained using other standard methodologies. Finally, as an additional robustness check, we re-estimated the product, firm and sector elasticities using 4-digit HS categories as our definition of sectors instead of 2-digit HS categories. We find a similar pattern of results, with a somewhat larger median product elasticity of 5.2, a median firm elasticity of 2.6, and a sector elasticity of 1.7. As discussed in Section 5.1 of the paper and reported in further detail in Section A.5.3 of this web appendix, we also demonstrate the robustness of our results to undertaking a grid search over the range of plausible values for the elasticity of substitution across firms and products.

<table>
<thead>
<tr>
<th>Percentile</th>
<th>Elasticity Across Products ($\sigma^U_{s\bar{u}}$)</th>
<th>Elasticity Across Firms ($\sigma^F_{s\bar{u}}$)</th>
<th>Elasticity Across Sectors ($\sigma^G_{s\bar{u}}$)</th>
<th>Product-Firm Difference ($\sigma^U_{s\bar{u}} - \sigma^F_{s\bar{u}}$)</th>
<th>Firm-Sector Difference ($\sigma^F_{s\bar{u}} - \sigma^G_{s\bar{u}}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Min</td>
<td>4.34</td>
<td>1.80</td>
<td>1.69</td>
<td>1.36</td>
<td>0.11</td>
</tr>
<tr>
<td>5th</td>
<td>4.44</td>
<td>2.09</td>
<td>1.69</td>
<td>1.63</td>
<td>0.40</td>
</tr>
<tr>
<td>25th</td>
<td>4.63</td>
<td>2.40</td>
<td>1.69</td>
<td>2.06</td>
<td>0.71</td>
</tr>
<tr>
<td>50th</td>
<td>5.01</td>
<td>2.68</td>
<td>1.69</td>
<td>2.39</td>
<td>0.99</td>
</tr>
<tr>
<td>75th</td>
<td>5.54</td>
<td>3.02</td>
<td>1.69</td>
<td>2.82</td>
<td>1.34</td>
</tr>
<tr>
<td>95th</td>
<td>6.88</td>
<td>3.40</td>
<td>1.69</td>
<td>4.33</td>
<td>1.71</td>
</tr>
<tr>
<td>Max</td>
<td>8.47</td>
<td>4.14</td>
<td>1.69</td>
<td>4.43</td>
<td>2.45</td>
</tr>
</tbody>
</table>

Note: Estimated elasticities of substitution from the reverse-weighting estimator discussed in Section 3 of the paper and in Section A.3 of this web appendix. Sectors are 2-digit Harmonized System (HS) codes; firms correspond to foreign exported brands within each foreign country within each sector; and products $u$ reflect 8-digit HS codes within exported brands within sectors.

Table A.1: Estimated Elasticities of Substitution, Within Firms ($\sigma^U_{s\bar{u}}$), Across Firms ($\sigma^F_{s\bar{u}}$) and Across Sectors ($\sigma^G_{s\bar{u}}$) using Chilean Data

In Figure A.5.1, we plot our estimated product, firm and sector elasticities of substitution ($\hat{\sigma}^U_{s\bar{u}}$, $\hat{\sigma}^F_{s\bar{u}}$, $\hat{\sigma}^G_{s\bar{u}}$), sorted based on the ranking of the estimated firm elasticity of substitution ($\hat{\sigma}^F_{s\bar{u}}$). We also show 95% confidence intervals for the estimated product and firm elasticities of substitution ($\hat{\sigma}^U_{s\bar{u}}$, $\hat{\sigma}^F_{s\bar{u}}$) based on bootstrapped standard errors. As can be seen in the figure, we find a natural ordering where $\hat{\sigma}^U_{s\bar{u}} > \hat{\sigma}^F_{s\bar{u}} > \hat{\sigma}^G_{s\bar{u}}$. We also find that the confidence intervals are narrow enough such that the product elasticity is significantly larger than the firm elasticity ($\hat{\sigma}^U_{s\bar{u}} > \hat{\sigma}^F_{s\bar{u}}$) at the 5 percent level of significance for 98 percent of sectors, and the firm elasticity is significantly larger than the sector elasticity ($\hat{\sigma}^F_{s\bar{u}} > \hat{\sigma}^G_{s\bar{u}}$) at this significance level for 88 percent of sectors.
Figure A.6.1: Estimated Elasticities of Substitution, Within Firms ($\hat{s}_U$), Across Firms ($\hat{s}_F$) and Across Sectors ($\hat{s}_G$), sorted based on the ranking of $\hat{s}_F$ (Chile Data)

### A.6.2 Exporter Price Indexes Across Sectors and Countries

We next show that we find a similar pattern of results for the exporter price indexes across countries and sectors using our Chilean data as using our U.S. data in Section 5.2 of the paper.

In the four panels of Figure A.6.2, we display the log of the exporter price index ($\ln P_{jigt}$) against its components using the Chilean data, where each observation is an exporting country and sector pair. For brevity, we show results for 2014, but find the same pattern for the other years in our sample. Whereas we show bin scatters using the U.S. data in Figure 2 in the paper, we show the observations for each exporting country and sector using our Chilean data in Figure A.6.2. In the top left panel, we compare the log exporter price index ($\ln P_{jigt}$) to average log product prices ($\ln P_{ut}$). In the special case in which firms and products are perfect substitutes within sectors ($\sigma_U = \sigma_F = \infty$) and there are no differences in demand/quality ($\varphi_{ft} = \varphi_{mt}$ for all $f, m$ and $\varphi_{ut} = \varphi_{lt}$ for all $u, l$), these two variables would be perfectly correlated. In contrast to these predictions, we find a positive but imperfect correlation, with an estimated regression slope of 0.24 and $R^2$ of essentially zero. In other words, average prices are weakly correlated with the true CES price index, which underscores the problem of using average prices as a proxy for the CES price index.

In the remaining panels of Figure A.6.2, we explore the three sources of differences between the exporter price index and average log product prices. As shown in the top-right panel, exporter-sectors with high average prices (horizontal axis) also have high average demand/quality (vertical axis), so that the impact of
Figure A.6.2: Exporter-Sector Price Indexes and their Components Versus Average Log Product Prices, 2014 (Chilean data)
higher average prices in raising sourcing costs is partially offset by higher average demand/quality. This positive relationship between average demand/quality and prices is strong and statistically significant, with an estimated regression slope of 0.59 and $R^2$ of 0.34. This finding of a tight connection between higher demand and higher prices is consistent with the quality interpretation of demand stressed in Schott (2004), in which producing higher quality incurs higher production costs.3

In the bottom-left panel of Figure A.6.2, we show that the contribution from the number of varieties to the exporter-sector price index exhibits an inverse U-shape, at first increasing with average prices before later decreasing. This contribution ranges by more than six log points, confirming the empirical relevance of consumer love of variety. In contrast, in the bottom-right panel of Figure 2, we show that the contribution from the dispersion of demand-adjusted prices displays the opposite pattern of a U-shape, at first decreasing with average prices before later increasing. While the extent of variation is smaller than for the variety contribution, this term still fluctuates by more than four log points between its minimum and maximum value. Therefore, the imperfect substitutability of firms and products implies important contributions from the number of varieties and the dispersion in demand-adjusted prices across those varieties towards the true cost of sourcing goods across countries and sectors.

These non-conventional determinants are not only important in the cross-section but are also important for changes in the cost of sourcing goods over time. A common empirical question in macroeconomics and international trade is the effect of price shocks in a given sector and country on prices and real economic variables in other countries. However, it is not uncommon to find that measured changes in prices often appear to have relatively small effects on real economic variables, which has stimulated research on “elasticity puzzles” and “exchange rate disconnect.” Although duality provides a precise mapping between prices and quantities, the actual price indexes used by researchers often differ in important ways from the formulas for price indexes from theories of consumer behavior. For example, as discussed in the paper, our average price term is the log of the “Jevons Index,” which is used by the U.S. Bureau of Labor Statistics (BLS) as part of its calculation of the consumer price index. Except in special cases, however, this average price term will not equal the theoretically-correct measure of the change in the unit expenditure function.

We first demonstrate this point for aggregate import prices. In Figure A.6.3, we use equation (27) in the paper to decompose the log change in aggregate import price indexes ($E_{jt}^T \left[ \Delta \ln P_{jt}^G \right]$) for Chile from 2008-14, where the analogous results for the U.S. are reported in the third column of Table 2 in the paper. This figure provides some important insights into why it is difficult to link import behavior to conventional price measures. If one simply computed the change in the cost of imported goods using a conventional Jevons Index of the prices of those goods (the first term in equation (27) in the paper), one would infer a substantial increase in the cost of imported goods of around 9.2 percent over this time period (prices are measured in current price U.S. dollars). However, this positive contribution from higher prices of imported goods was offset by a substantial negative contribution from firm entry (variety). This expansion in firm variety reduced the cost of imported goods by around 11.7 percent. By contrast, country-sector and firm dispersion fell over

3This close relationship between demand/quality and prices is consistent the findings of a number of studies, including the analysis of U.S. barcode data in Hottman et al. (2016) and the results for Chinese footwear producers in Roberts et al. (2011).
this period, which raised the cost of imported goods, and offset some of the variety effects. As a result, the true increase in aggregate import prices from 2008-14 was only 4.4 percent, less than half of the value implied by a conventional Jevons Index. In other words, the true measure of aggregate import prices is strongly affected by factors other than movements in average prices.

We next show that this point applies not only to aggregate import prices but also to changes in the exporter price indexes $\Delta \ln P_{jigt}^E$, that summarize the cost of sourcing goods across countries and sectors. Figure A.6.4 displays the same information as in Figure A.6.2, but for log changes from 2008-2014 rather than for log levels in 2014 (where the corresponding results using the U.S. data are in Figures 2 and 3 respectively in the paper). Whereas we show bin scatters using the U.S. data in the paper, we again show the observations for each exporting country and sector using our Chilean data in this web appendix. In changes, the correlation between average prices and the true model-based measure of the cost of sourcing goods is even weaker and the role for demand/quality is even greater. Indeed, the slope for the regression of average log changes in demand/quality on average log changes in prices is almost one, indicating that most price changes are almost completely offset by demand/quality changes. As in the U.S. data, this result indicates a problem for standard price indexes that assume no demand or quality shifts for commonly available goods, such as the Sato-Vartia price index.
Figure A.6.4: Log Changes in Sector-Exporter Price Indexes and their Components Versus Average Log Changes in Product Prices, 2008-2014 (Chilean data)
A.6.3 Trade Patterns

The similarity of our findings for exporter price indexes for Chile and the U.S. suggests that we should also find similar results for patterns of trade, because revealed comparative advantage (RCA) depends on relative price indexes. In this section of the web appendix, we confirm that this is indeed the case.

In Table A.2, we present the decompositions of RCA from equation (31) in Section 5.3 of the paper, but using our Chilean data instead of our U.S. data (see Table 3 in the paper for the U.S. results). In Columns (1)-(2), we report results for levels of RCA. In Columns (3) and (4), we present the corresponding results for changes in RCA. While Columns (1) and (3) undertake these decompositions down to the firm level, Columns (2) and (4) undertake them all the way down to the product-level. For brevity, we concentrate on the results of the full decomposition in Columns (2) and (4). We find that average prices are comparatively unimportant in explaining patterns of trade. In the cross-section, average product prices account for 12.6 percent of the variation in RCA. In the time-series, we find that average prices are even less important, accounting for only 9 percent of the variation. These results reflect the low correlations between average prices and exporter price indexes seen in the last section. If average prices are weakly correlated with exporter price indexes, they are unlikely to matter much for RCA, because RCA is determined by relative exporter price indexes. By contrast, we find that average demand/quality is two to three times more important than average prices, with a contribution of 23 percent for the levels of RCA and 36 percent for the changes in RCA.

By far the most important of the different mechanisms for trade in Table A.2 is firm variety, which accounts for 34 and 46 percent of the level and change of RCA respectively. We also find a substantial contribution from the dispersion of demand-adjusted prices across firms, particularly in the cross-section, where this term accounts for 30 percent of the variation in RCA. In the time-series, changes in the dispersion of demand-adjusted prices across common firms are relatively less important, although they still account for 9 percent of the changes in RCA. On the one hand, our findings for firm variety are consistent with research that emphasizes the role of the extensive margin in understanding patterns of trade (e.g. Hummels and Klenow 2005, Chaney 2008). On the other hand, our findings for the dispersion of demand-adjusted prices across common varieties imply that the intensive margin is also important (consistent with the analysis for a log normal distribution in Fernandes et al. 2015). In particular, we find quantitatively relevant differences in the second moment of the distribution of demand-adjusted prices across common products and firms within exporters and sectors.

We now show that the non-conventional forces of variety, average demand/quality and the dispersion of demand-adjusted prices are also important for understanding movements in aggregate Chilean imports from its largest trade partners, consistent with our U.S. results in Section 5.3 of the paper. In Figure A.6.5, we show the time-series decompositions of aggregate import shares from equation (25) in the paper for Chile’s top-six trade partners. As apparent from the figure, we can account for the substantial increase in China’s market share over the sample period by focusing mostly on increases in firm variety (orange), average firm demand/quality (gray), and the dispersion of demand-adjusted prices across firms (light blue).4 In contrast,

4Our finding of an important role for firm entry for China is consistent with the results for export prices in Amiti, Dai, Feenstra, and Romalis (2016). However, their price index is based on the Sato-Vartia formula, which abstracts from changes in demand/quality
<table>
<thead>
<tr>
<th></th>
<th>Log Level RCA 2014</th>
<th>Log Change RCA 2008-14</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Firm-Level</td>
<td>Product-Level</td>
</tr>
<tr>
<td>Firm Price Index</td>
<td>0.126</td>
<td>-</td>
</tr>
<tr>
<td>Firm Demand</td>
<td>0.233</td>
<td>0.233</td>
</tr>
<tr>
<td>Firm Variety</td>
<td>0.344</td>
<td>0.344</td>
</tr>
<tr>
<td>Firm Dispersion</td>
<td>0.297</td>
<td>0.297</td>
</tr>
<tr>
<td>Product Prices</td>
<td>-</td>
<td>0.107</td>
</tr>
<tr>
<td>Product Variety</td>
<td>-</td>
<td>0.013</td>
</tr>
<tr>
<td>Product Dispersion</td>
<td>-</td>
<td>0.010</td>
</tr>
</tbody>
</table>

Note: Variance decomposition for the log level of RCA in 2014 and the log change in RCA from 2008-14 (from equation (31) in the paper).

Table A.2: Variance Decomposition Chilean RCA

![Figure A.6.5: Country Aggregate Shares of Chilean Imports](image-url)
average product prices (green) increased more rapidly for China than for the other countries in our sample, which worked in the opposite direction to reduce China’s market share. In other words, our decomposition indicates that the reason for the explosive growth of Chinese exports was not due to cheaper Chinese exports, but rather substantial firm entry (variety), product upgrading (demand/quality), and improvements in the performance of leading firms relative to lagging firms (the dispersion of demand-adjusted prices). By contrast the dramatic falls in import shares from Argentina and Brazil were driven by a confluence of factors that all pushed in the same direction: higher average product prices, firm exit (variety), a deterioration in the performance of leading firms relative to lagging firms (the dispersion of demand-adjusted prices), and falls in average demand/quality relative to other countries.

A.6.4 Additional Theoretical Restrictions

We now compare our approach, which exactly rationalizes both micro and macro trade data, with special cases of this approach that impose additional theoretical restrictions. We show that we find a similar pattern of results using the Chilean data as using the U.S. data in Section 5.4 of the paper. As a result of imposing additional theoretical restrictions, these special cases no longer exactly rationalize the micro trade data, and we quantify the implications of these departures from the micro data for macro trade patterns and prices.

Almost all existing theoretical research with CES demand in international trade is encompassed by the Sato-Vartia price index, which assumes no shifts in demand/quality for common varieties. Duality suggests that there are two ways to assess the importance of this assumption. First, we can work with a price index and examine how a CES price index that allows for demand shifts (i.e., the UPI in equation (16) in the paper) differs from a CES price index that does not allow for demand shifts (i.e., the Sato-Vartia index). Since the common goods component of the UPI (CG-UPI) and the Sato-Vartia indexes are identical in the absence of demand shifts, the difference between the two is a metric for how important demand shifts are empirically. Second, we can substitute each of these price indexes into our expression for revealed comparative advantage (RCA) in equation (22) in the paper, and examine how important the assumption of no demand shifts is for understanding patterns of trade. Because we know that the UPI perfectly rationalizes the data, any deviation from the data arising by using a different price index must reflect the effect of the restrictive assumptions used in the index’s derivation. In order to make the comparison fair, we need to also adjust the Sato-Vartia index for variety changes, which we do by using the Feenstra (1994) index, which is based on the same no-demand-shifts assumption for common goods, but adds the variety correction term given in equation (16) in the paper to incorporate entry and exit.

In Figure A.6.6, we report the results of these comparisons using our Chilean data, which corresponds to Figure 5 in the paper using our U.S. data. The top two panels consider exporter price indexes, while the bottom two panels examine RCA. In the top-left panel, we compare the Sato-Vartia exporter price index (on the vertical axis) with our common goods exporter price index (the CG-UPI on the horizontal axis), where each observation is an exporter-sector pair. If the assumption of time-invariant demand/quality were satisfied in the data, these two indexes would be perfectly correlated with one another and aligned on the 45-degree
Again, we find little relationship between them. The reason is immediately apparent if one recalls the top-right panel of Figure A.6.4, which shows that price shifts are strongly positively correlated with demand shifts. The Sato-Vartia price index fails to take into account that higher prices are typically offset by higher demand/quality. In the top-right panel, we compare the Feenstra exporter price index (on the vertical axis) with our overall exporter price index (the UPI on the horizontal axis), where each observation is again an exporter-sector pair. These two price indexes have exactly the same variety correction term, but use different common goods price indexes (the CG-UPI and Sato-Vartia indexes respectively). The importance of the variety correction term as a share of the overall exporter price index accounts for the improvement in the fit of the relationship. However, the slope of the regression line is only around 0.5, and the regression $R^2$ is about 0.1. Therefore, the assumption of no shifts in demand/quality for existing goods results in substantial deviations between the true and measured costs of sourcing goods from an exporter and sector.

In the bottom left panel, we compare predicted changes in RCA based on relative exporter Sato-Vartia price indexes (on the vertical axis) with actual changes in RCA (on the horizontal axis). As the Sato-Vartia price index has only a weak correlation with the UPI, we find that it has little predictive power for changes in RCA, which are equal to relative changes in the UPI across exporters and sectors. Hence, observed changes in trade patterns are almost uncorrelated with the changes predicted under the assumption of no shifts in demand/quality and no entry and exit of firms and products. In the bottom right panel, we compare predicted
changes in RCA based on relative exporter Feenstra price indexes (on the vertical axis) with actual changes in RCA (on the horizontal axis). The improvement in the fit of the relationship attests to the importance of adjusting for entry and exit. However, again the slope of the regression line is only around 0.5 and the regression $R^2$ is about 0.1. Therefore, even after adjusting for the shared entry and exit term, the assumption of no demand shifts for existing goods can generate predictions for changes in trade patterns that diverge substantially from those observed in the data. This importance of changes in demand/quality for surviving goods for the evolution of RCA again highlights the role of this mechanism in understanding the churning in trade patterns documented in Hanson, Lind and Muendler (2016). To the extent that these changes in demand/quality are driven by endogenous investments in innovation, these findings are also consistent with model in which comparative advantage arises endogenously because of product and process innovation, as in Grossman and Helpman (1991).

Although the Sato-Vartia price index assumes no shifts in demand/quality for surviving varieties, it makes no functional form assumptions about the cross-sectional distributions of prices, demand/quality and expenditure shares. We now examine the implications of imposing additional theoretical restrictions on these cross-sectional distributions. In particular, an important class of existing trade theories assumes not only a constant demand-side elasticity but also a constant supply-side elasticity, as reflected in the assumption of Fréchet or Pareto productivity distributions. As our approach uses only demand-side assumptions, we can examine the extent to which these additional supply-side restrictions are satisfied in the data. In particular, we compare the observed data for firm sales and our model solutions for the firm price index and firm demand/quality ($\ln V_{ft} \in \{\ln X_{ft}, \ln P_{ft}, \ln \varphi_{ft}\}$) with their theoretical predictions under alternative supply-side distributional assumptions.

To derive these theoretical predictions, we use the QQ estimator. The QQ estimator compares the empirical quantiles in the data with the theoretical quantiles implied by alternative distributional assumptions. As shown in Section A.5.4 of this web appendix, under the assumption that a firm variable $V_{ft}$ has an untruncated Pareto distribution, we obtain the following theoretical prediction for the quantile of the logarithm of that variable:

$$\ln \left( V_{ft}^F \right) = \ln V_{ft}^F - \frac{1}{a_g^V} \ln \left[ 1 - F_{jigt} \left( V_{ft}^F \right) \right].$$  \hspace{1cm} (A.6.1)

where $F_{jigt} (\cdot)$ is the cumulative distribution function; $\ln V_{jigt}^F$ is the lower limit of the support of the untruncated Pareto distribution, which is a constant across firms $f$ for a given importer $j$, exporter $i$, sector $g$ and year $t$; $a_g^V$ is the shape parameter of this distribution, which we allow to vary across sectors $g$.

We estimate equation (A.6.1) by OLS using the empirical quantile for $\ln \left( V_{ft}^F \right)$ on the left-hand side and the empirical estimate of the cumulative distribution function for $F_{jigt} \left( V_{ft}^F \right)$ on the right-hand side. We estimate this regression for each sector across foreign firms (allowing the slope coefficient $a_g^V$ to vary across sectors) and including fixed effects for each exporter-sector-year combination (allowing the intercept $\ln V_{jigt}^F$ to vary across exporters, sectors and time). The fitted values from this regression correspond to the predicted theoretical quantiles, which we compare to the empirical quantiles observed in the data. Under the null hypothesis of a Pareto distribution, there should be a linear relationship between the theoretical and empirical
quantiles that coincides with the 45-degree line.

In Figure A.6.7, we show the predicted theoretical quantiles (vertical axis) against the empirical quantiles (horizontal axis) using our Chilean data. We display results for log firm imports (top left), log firm price indexes (top right) and log firm demand/quality (bottom left). In each case, we observe sharp departures from the linear relationship implied by a Pareto distribution, with the actual values below the predicted values in both the lower and upper tails. Following the same approach as in Section 5.4 of the paper, we estimate the regression in equation (A.6.1) separately for observations below and above the median, and compare the estimated coefficients. Consistent with the U.S. results in Figure A.5.4 of this web appendix, we find substantial departures from linearity using the Chilean data, which are statistically significant at conventional levels.

As a point of comparison, we also examine the alternative distributional assumption of a log normal distribution. As shown in Section A.5.4 of this web appendix, under this distributional assumption, we obtain the following theoretical prediction for the quantile of the logarithm of a variable $V_{ft}^F$:

$$
\ln \left( V_{ft}^F \right) = \kappa_{jigt}^V + \chi_g^V \Phi^{-1} \left( F_{jigt} \left( V_{ft}^F \right) \right),
$$

(A.6.2)

where $\Phi^{-1} (\cdot)$ is the inverse of the normal cumulative distribution function; $\kappa_{jigt}^V$ and $\chi_g^V$ are the mean and standard deviation of the log variable, such that $\ln \left( V_{ft}^F \right) \sim \mathcal{N} \left( \kappa_{jigt}^V, \left( \chi_g^V \right)^2 \right)$; we make analogous assumptions about these parameters as for the untruncated Pareto distribution above; we allow the parameter controlling the mean ($\kappa_{jigt}^V$) to vary across exporters $i$, sectors $g$ and time $t$ for a given importer $j$; we allow the parameter controlling dispersion ($\chi_g^V$) to vary across sectors $g$.

Again we estimate equation (A.6.2) by OLS using the empirical quantile for $\ln \left( V_{ft}^F \right)$ on the left-hand side and the empirical estimate of the cumulative distribution function for $F_{jigt} \left( V_{ft}^F \right)$ on the right-hand side. We estimate this regression for each sector across foreign firms (allowing the slope coefficient $\chi_g^V$ to vary across sectors) and including fixed effects for each exporter-sector-year combination (allowing the intercept $\kappa_{jigt}^V$ to vary across exporters, sectors and time).

In Figure A.6.8, we show the predicted log normal theoretical quantiles (vertical axis) against the empirical quantiles (horizontal axis) using our Chilean data. Again we display results for log firm imports (top left), log firm price indexes (top right) and log firm demand/quality (bottom left). In each case, we find that the relationship between the theoretical and empirical quantiles is closer to linearity for a log-normal distribution than for a Pareto distribution, which is consistent with Bas, Mayer and Thoenig (2017). Nonetheless, we observe substantial departures from the theoretical predictions of a log-normal distribution, and we reject the null hypothesis of normality at conventional levels of significance for the majority of sectors using a Shapiro-Wilk test. Following the same approach as in Section 5.4 of the paper, we also estimate the regression in equation (A.6.2) separately for observations below and above the median, and compare the estimated coefficients. Consistent with the U.S. results in Figure A.5.4 of this web appendix, we again find substantial departures from linearity using the Chilean data, which are statistically significant for the majority of sectors at conventional levels.
Figure A.6.7: Theoretical and Empirical Quantiles for Chile (Pareto Distribution)
Figure A.6.8: Theoretical and Empirical Quantiles for Chile (Log Normal Distribution)
A.6.5 Additional Reduced-Form Evidence

In Figures A.6.9-A.6.12, we confirm that our Chilean trade transaction data have the same reduced-form properties as our U.S. data and as found in other empirical studies using international trade transactions data (see for example Bernard, Jensen and Schott 2009 and Bernard, Jensen, Redding and Schott 2009 for the U.S.; Mayer, Melitz and Ottaviano 2014 for France; and Manova and Zhang 2012 for China).

First, Chilean imports are highly concentrated across countries and characterized by a growing role of China over time. As shown in Figure A.6.9, Chile’s six largest import sources in 2007 were (in order of size) China, the U.S., Brazil, Germany, Mexico, and Argentina, which together accounted for more than 60 percent of its imports. Between 2007 and 2014, China’s import share grew by over 50 percent, with all other major suppliers except Germany experiencing substantial declines in their market shares.

Second, we find high rates of product and firm turnover and evidence of selection conditional on product and firm survival. In Figure A.6.10, we display the fraction of firm-product observations and import value by tenure (measured in years) for 2014, where recall that firms here correspond to foreign exporting firms. Around 50 percent of the firm-product observations in 2014 have been present for one year or less, but the just over 10 percent of these observations that have survived for at least seven years account for over 40 percent of import value.

Third, we find that international trade is dominated by multi-product firms. In Figure A.6.11, we display the fraction of firm observations and import value in 2014 accounted for by firms exporting different numbers of products. Although less than 30 percent of exporting firms are multi-product, they account for more than 70 percent of import value.

Fourth, we find that the extensive margins of firm and product exporting account for most of the cross-section variation in aggregate trade. In Figure A.6.12, we display the log of the total value of Chilean imports from each foreign country, the log number of firm-product observations with positive trade for that country, and the log of average imports per firm-product observation with positive trade from that country. We display these three variables against the rank of countries in Chile’s total import value, with the largest country assigned a rank of one (China). By construction, total import value falls as we consider countries with higher and higher ranks. Substantively, most of this decline in total imports is accounted for by the extensive margin of the number of firm-product observations with positive trade, whereas the intensive margin of average imports per firm-product observation with positive trade remains relatively flat.

Therefore, across these and a range of other empirical moments, the Chilean data are representative of empirical findings using international trade transactions data for a number of other countries.
Figure A.6.9: Country Shares of Chilean Imports over Time

Note: Data are for 2014. Tenure is the number of years a firm-product observation has existed since 2007. Number of observations 947773

Figure A.6.10: Distribution Firm-Product Observations and Import Value by Tenure 2014 (Chilean Data)
Figure A.6.11: Distribution of Firm Observations Across Number of Products 2014 (Chilean Data)

Figure A.6.12: Extensive and Intensive Margins of Firm-Product Imports Across Countries 2014 (Chilean Data)
A.7 Unobserved Differences in Product Composition

In this section of the web appendix, we show that our approach allows for unobserved differences in composition within observed product categories, which enter the model in the same way as unobserved differences in demand/quality for each observed product category. In the paper, we assume for simplicity that the products supplied by firms are the same as those observed in the data, which enables us to abstract from these unobserved differences in product composition. We now generalize our results to the case in which firms supply products at a more disaggregated level (e.g. unobserved barcodes) than the categories observed in the data (Harmonized System (HS) categories).

A.7.1 True Data Generating Process

We suppose that the true data generating process is as follows. At the aggregate level, we have sectors \( g \); below sectors we have firms \( f \); below firms we have products \( u \); and below products we have barcodes \( b \). Aggregate utility and the consumption index for each sector remain unchanged. The consumption index for each firm \((C_{ft})\) is defined over an unobserved consumption index for each product \((C_{ut}^U)\):

\[
C_{ft}^F = \left[ \sum_{u \in \Omega_{ft}^U} \left( \frac{\varrho_{ut}^U}{\varphi_{ut}^U} \right)^{\frac{1}{\sigma_g^U}} \right]^{\frac{1}{\sigma_g^U}}, \quad \sigma_g^U > 1, \varrho_{ut}^U > 0, \tag{A.7.1}
\]

where \(\sigma_g^U\) is the elasticity of substitution across products within the firm; \(\varrho_{ut}^U\) is the demand/quality for each product; and \(\Omega_{ft}^U\) is the set of products supplied by firm \(f\) at time \(t\). Each product consumption index \((C_{ut}^U)\) is defined over the unobserved consumption of each barcode \((C_{bt}^B)\):

\[
C_{ut}^U = \left[ \sum_{b \in \Omega_{bt}^B} \left( \frac{\varphi_{bt}^B}{\varrho_{bt}^B} \right)^{\frac{1}{\sigma_g^B}} \right]^{\frac{1}{\sigma_g^B}}, \quad \sigma_g^B > 1, \varphi_{bt}^B > 0. \tag{A.7.2}
\]

Similarly, the dual price index for each firm \((P_{ft}^F)\) is defined over an unobserved dual price index for each product \((P_{ut}^U)\):

\[
P_{ft}^F = \left[ \sum_{u \in \Omega_{ft}^U} \left( \frac{P_{ut}^U}{\varphi_{ut}^U} \right)^{1-\sigma_g^U} \right]^{\frac{1}{1-\sigma_g^U}}, \tag{A.7.3}
\]

and this unobserved dual price index for each product \((P_{ut}^U)\) is defined over the unobserved price of each barcode \((P_{bt}^B)\):

\[
P_{ut}^U = \left[ \sum_{b \in \Omega_{bt}^B} \left( \frac{P_{bt}^B}{\varrho_{bt}^B} \right)^{1-\sigma_g^B} \right]^{\frac{1}{1-\sigma_g^B}}. \tag{A.7.4}
\]
A.7.2 Observed Data

Suppose that in the data we observe the total value of sales of each product \((E_{ut}^U)\), which corresponds to the sum of the sales of all the unobserved barcodes \((E_{ut}^B)\):

\[
E_{ut}^U = P_{ut}^U C_{ut}^U = \sum_{b \in \Omega_{ut}^B} E_{bt}^B = \sum_{b \in \Omega_{ut}^B} P_{bt}^B C_{bt}^B.
\]  

We also observe the total physical quantity of each product \((Q_{ut}^U)\), which corresponds to the sum of the physical quantities of all barcodes \((Q_{ut}^B)\). Dividing sales by quantities for each product, we can compute a unit value for each product \((P_{ut}^U)\):

\[
P_{ut}^U = P_{ut}^U Q_{ut}^U = E_{ut}^U.
\]

which implies that the ratio of observed unit values to unobserved price indexes is the inverse of the ratio of observed physical quantities to unobserved consumption indexes:

\[
\frac{P_{ut}^U}{P_{ut}^B} = \frac{1}{Q_{ut}^U / C_{ut}^U}.
\]

A.7.3 Relationship Between Observed and Unobserved Variables

We now use these relationships to connect the observed physical quantities and unit values \((Q_{ut}^U, P_{ut}^U)\) to the true unobserved consumption and price indexes \((C_{ft}^F, P_{ft}^F)\). The firm consumption index \((C_{ft}^F)\) can be re-written in terms of the observed physical quantities of each product \((Q_{ut}^U)\) and a quality-adjustment parameter \((\theta_{ut}^U)\) that captures the demand/quality of each product \((q_{ut}^U)\) and the discrepancy between the observed quantity of each product \((Q_{ut}^U)\) and the unobserved product consumption index \((C_{ut}^U)\):

\[
C_{ft}^F = \left[ \sum_{u \in \Omega_{ft}^U} \left( \theta_{ut}^U Q_{ut}^U \left( \frac{Q_{ut}^U}{Q_{ut}^U} \right)^{\frac{c_{ut}^U - 1}{c_{ut}^U}} \right) \right]^{\frac{c_{ut}^U}{c_{ut}^U - 1}},
\]

where the quality-adjustment parameter is defined as:

\[
\theta_{ut}^U = \phi_{ut}^U C_{ut}^U Q_{ut}^U.
\]

Combining this definition in equation (A.7.9) with the relationship between observed and unobserved variables in equation (A.7.7), the firm price index \((P_{ft}^F)\) also can be re-written in terms of the observed unit values for each product \((P_{ut}^U)\) and this same quality-adjustment parameter \((\theta_{ut}^U)\):

\[
P_{ft}^F = \left[ \sum_{u \in \Omega_{ft}^U} \left( \frac{P_{ut}^U}{\theta_{ut}^U} \right)^{1 - c_{ut}^U} \right]^{\frac{1}{1 - c_{ut}^U}}.
\]
Note that equations (A.7.8) and (A.7.10) are identical to equations (A.7.1) and (A.7.3), except that the unobserved consumption and price indexes \((C_{ft}, P_{ft})\) in equations (A.7.1) and (A.7.3) are replaced by the observed quantities and unit values \((Q_{ut}, P_{ut})\), and the unobserved demand/quality parameters \((\theta_{ut}^{l})\) are replaced by the quality-adjustment parameter \((\theta_{ut}^{l})\). Therefore, we can implement our entire analysis using the observed quantities and unit values \((Q_{ut}, P_{ut})\) and the quality-adjustment parameter \((\theta_{ut}^{l})\). We cannot break out this quality-adjustment parameter \((\theta_{ut}^{l})\) into the separate contributions of true product quality \((\theta_{ut}^{j})\) and the discrepancy between the true consumption index and observed physical quantities \((C_{ut}/Q_{ut})\). But we can use our estimation procedure to estimate the elasticity of substitution across products \((\sigma_{g}^{J})\), recover the quality-adjustment parameter for each product \((\theta_{ut}^{l})\), recover the true firm consumption and price indexes \((C_{ft}^{F}, P_{ft}^{F})\), estimate the elasticity of substitution across firms \((\sigma_{g}^{F})\), and implement the remainder of our analysis.
References


