Outsourcing and Volatility

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Abstract:
This paper studies the second-moment properties of global outsourcing. It documents a new empirical regularity: outsourcing industries in Mexico experience fluctuations in employment that are twice as volatile as the corresponding industries in the U.S. The paper develops a theoretical model of outsourcing to explain this stylized fact. The model features an extensive margin of outsourcing that responds endogenously to shocks in demand, and transmits those shocks across borders.

JEL classification: F1
I. Introduction

Foreign outsourcing, the arrangement whereby firms contract with independent counterparts in another country to carry out particular stages of production, has grown over the last fifteen years to become an important part of the trade relationship between the U.S. and Mexico. It is also of growing importance for trade between the E.U. and emerging economies in Europe, and in global trade with China. In Mexico, employment in export assembly plants engaged in outsourcing grew ten-fold from 0.12 million in 1980 to 1.2 million in 2006. The sector accounts for just under 3% of Mexico’s total GDP, 20% of Mexican manufacturing value added, and nearly half of the country’s exports. While Mexican officials have hailed the export assembly plants for their contribution to economic growth, some have also complained that the sector is fickle and subject to excessive volatility.\(^1\) The assembly plants, known as *maquiladoras*, are seen as a channel by which the U.S. exports to Mexico a portion of its employment fluctuations over the business cycle. This phenomenon is beginning to be investigated in the literature on real business cycles across countries. Our paper is unique in focusing on the maquiladora sector, using a model that emphasizes the high degree of substitution between production in those plants and their U.S. counterparts.\(^2\)

We begin by documenting the variance in outsourcing industries in Mexico. These industries are composed of maquiladoras to which U.S. and other foreign firms outsource the assembly of inputs into final outputs. Our data cover Mexico’s four largest outsourcing industries, which together account for three quarters of outsourcing production in the country: apparel, transportation equipment, computers and electronics, and electrical machinery. We match these industries to their counterparts in the United States. Our main empirical result is that in all four outsourcing industries

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\(^1\) See for example the news account of how the Mexican car industry is highly susceptible to fluctuations in demand for American brand automobiles in Dickerson (2005).

\(^2\) Burstein, Kurz, and Tesar (2008) develop a dynamic model of trade in intermediate inputs whose respective outputs are assumed to be complements in production, whereas we model international production in terms of outsourcing of variable cost activities that are highly substitutable. Our formulation draws on the outsourcing model in Feenstra and Hanson (1996, 1997). For other work on intermediate inputs and business cycle synchronization see Kose and Yi (2001, 2006) and Ambler, Cardia, and Zimmerman (2002).
the volatility of economic activity in Mexico is significantly higher than in the U.S.; averaging over
the four industries, volatility in Mexico is twice as high as in the U.S. One conjecture might be that
this simply reflects higher volatility in the Mexican economy overall. That conjecture is not true for
employment volatility, which is lower in Mexico. A second conjecture might be that higher volatility
in Mexico reflects the smaller size of industries in Mexico. However, our results are robust to
comparing Mexican industries with the more similarly sized industries of U.S. border states.³

To explain differential volatility in countries engaged in outsourcing, we develop a model of
global production sharing where the outsourcing decisions of firms respond to macroeconomic
shocks. The model relies on a continuum of products in the outsourcing sector, and for each product
an endogenous number of varieties. This structure combines the Dornbusch-Fisher-Samuelson (1977)
framework with the monopolistic competition model, as also done by Romalis (2004). Production in
the outsourcing sector requires two activities: a fixed-cost activity that occurs in the high-wage home
country, representing headquarter and managerial costs, and a variable-cost activity representing
assembly work that can be done at home or outsourced to a low-wage foreign country.

A key feature of the model is that the point along the product continuum at which firms in the
home country begin to outsource the variable-cost activity to the foreign country is endogenously
determined as firms compare the unit labor costs across borders. When the home country has a boom
in demand, the fact that home wages tend to be procyclical alters the outsourcing decision of some

³ A third conjecture might be that labor-market institutions differ between the countries, such that it is easier to hire
and fire employees in Mexico. However, when Botero et al. (2004) rank countries in terms of job security laws
restricting the hiring and firing of workers, Mexico ties for the most regulated among the 85 countries in the sample,
whereas the U.S. ranks as the fifth least regulated economy in this respect; these data are used in the analysis of
volatility and comparative advantage by Cunat and Melitz (2006). We also see this contrast in labor market
flexibility reflected in our employment data discussed below, in which volatility of employment in overall Mexican
manufacturing is lower than that for U.S. overall manufacturing. In Mexico, labor unions are active in maquiladoras
and other manufacturing establishments, both of which are subject to Mexican labor law (Otero and Pagan, 2002). In
practice, collective bargaining agreements in maquiladoras are seen as less restrictive than in the rest of Mexican
manufacturing, in that job classifications tend to be wider and firms tend to have greater leeway in establishing pay-
for-performance schemes (Fairris, 2003). At the same time, the decline in unionization in Mexican manufacturing,
which has been ongoing since the 1980s, is less evident among maquiladoras. The net effect of more flexible
collective bargaining agreements but more stable union presence appears to be that differences in labor relations
between maquiladoras and other Mexican manufacturing firms are relatively small.
firms. Since home workers become relatively more expensive to hire, firms that previously had not outsourced any production now find it profitable to do so. This shift in the extensive margin acts as a powerful mechanism for the international transmission of shocks, whereby U.S. producers shift unusually high levels of production abroad during a domestic economic boom, and the reverse during a recession. Even when the shock is a purely domestic one, the simulation shows that it is amplified in its transmission abroad, so that it has a greater impact on the outsourcing industries in the low-wage foreign country than on the domestic counterpart industries.

There are two reasons in our model why the shocks to home demand are amplified in their effect abroad. First, the foreign outsourcing sector, representing maquiladoras in Mexico, has smaller total employment than in the U.S. Any given shift in employment therefore has an amplified effect on foreign volatility. Second, as just noted, home wages are procyclical with home demand shocks. A home demand shock that raises demand in the outsourcing industry will tend to raise employment disproportionately in the foreign country, since outsourcing activity is shifted in that direction at the same time that employment is growing due to increased demand. In contrast, the shift in outsourcing activities offsets the employment increase at home.

The outsourcing sector is embedded in a two country, general equilibrium trade model, which also includes a homogeneous traded good in each country. Numerical examples, by way of stochastic simulation under demand and supply shocks, indicate that the mechanism provides a potential explanation for the extra volatility in Mexican outsourcing. The next section presents the data and empirical results. Section 3 presents the theoretical model, and section 4 discusses simulation results. Conclusions are given in section 5.

II. Data and Empirical Results

A. The Maquiladora Sector

Outsourcing by the U.S. to Mexico generally takes the form of U.S. firms producing parts and
components, exporting these intermediate inputs to Mexico to be assembled or processed into final goods, and re-importing the finished products. U.S. firms tend to specialize in R&D, component production, marketing, and other headquarters activities, while Mexican plants – the maquiladoras – tend to specialize in assembly services.\(^4\) Mexico is among the most important locations for global outsourcing by U.S. firms and the U.S. is by far and away Mexico’s largest trading partner. Over the period 2000-2003, the United States was the source country for 73.4% of the inputs imported by maquiladoras in Mexico and maquiladora exports back to the United States were equal to 5.3% of U.S. industry shipments (U.S. International Trade Commission, 2005).\(^5\) Maquiladoras have become an integral part of the Mexican economy, with their share of national manufacturing employment rising from 4.1% in 1980 to 24.5% in 2004 (Hanson, 2007).

Mexico first began to allow export assembly plants to operate in the country in the 1960s. The maquiladora sector did not reach an appreciable size until the government relaxed restrictions on inbound foreign investment in the 1980s. Initially, U.S. firms outsourcing to Mexico received favorable tariff treatment under the HS9802 provision of the U.S. tariff code (Feenstra, Hanson, and Swenson, 2000). Under HS9802, U.S. firms that manufacture components at home and have them processed into final goods abroad pay duties on the foreign value-added only when the goods are brought back into the United States. The North American Free Trade Agreement ended special tariff

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\(^4\) The Mexican government measures imports and exports by registered export-assembly plants in Mexico. Under Mexican trade policy, firms that export their output do not have to pay duties on any imported intermediate inputs used in production. To obtain duty-free status, a firm must register with the government. While under the North American Free Trade Agreement imports from the United States are not subject to duties in Mexico, imports from most other countries are. As a result, the vast majority of export-assembly plants in Mexico are registered. (Strictly speaking, export-assembly plants in Mexico may be registered either as maquiladoras or as PITEX (Program for Temporary Imports of Articles to be Exported) companies (see http://www.economia.gob.mx/). Only the former appear in our data. In 2003, maquiladora exports to the United States were 2.2 times those by PITEX companies.)

\(^5\) These figures apply to the four core outsourcing industries, described in the text. Comparing U.S. imports from Mexico to U.S. industry output may give a deceptive sense of the size of Mexico’s maquiladora sector relative to U.S. manufacturing. U.S. imports from Mexico include a substantial component of U.S. value added, in the form of the intermediate inputs produced in the United States and sent to Mexico for further processing. As an alternative measure of relative size, one might examine value added in the two countries. Over the period 2000-2003, the ratio of value added in Mexico’s maquiladoras to value added in U.S. manufacturing was 0.034 in the four core outsourcing industries, based on annual data (U.S. International Trade Commission, 2005).
treatment for U.S. firms outsourcing to Mexico. But, as Figure 1 shows, it did not slow growth in production sharing. Growth in real value added by maquiladoras accelerated after NAFTA was implemented, increasing by over 100 log points between 1994 and 2005. Far from removing the incentive for Mexico to specialize in assembly services, NAFTA freed resources Mexico had devoted to domestic production to move into export assembly.

Most outsourcing by U.S. firms in Mexico occurs in one of four industries: apparel, electronic accessories (including computer parts and electronic circuitry), electrical machinery (including televisions and other small domestic appliances), and transport equipment and parts (primarily motor vehicles). Over 1990 to 2005 these four industries accounted for three-quarters of employment in the maquiladora sector. These industries are amenable to global production sharing because firms need not perform all stages—R&D, component production, final assembly—in the same location, and the stages vary in their factor intensities, giving multinational firms an incentive to locate labor-intensive activities in low-wage countries.

**B. Volatility in Employment**

As a measure of volatility, we shall focus on the relative variance of employment in U.S. manufacturing industries and the plants to which they outsource in Mexico.\(^6\) We compute this variance using the monthly employment of production workers, which is available for both countries. We match Mexico’s four primary outsourcing industries (assembly of apparel items; electronic materials and accessories, including computer parts; assembly of electrical machinery and equipment; and construction and assembly of transport equipment and parts) with their closest U.S. three-digit industry matches (apparel manufacturing, NAICS 315; computer and electronic product manufacturing, NAICS 334; electrical equipment, appliance, and component manufacturing, NAICS 335; and transportation equipment manufacturing, NAICS 336). Data are available beginning in

\(^6\) We have confirmed that similar results hold when using the wage bill for production workers instead of employment; see our working paper (Bergin, Feenstra and Hanson, 2007).
However, in late 1994 there was a large depreciation of the peso, as capital fled Mexico, and in 1995 aggregate output dropped sharply. Given the exposure of the maquiladora sector to exchange-rate fluctuations, including the peso-crisis years in our sample could make volatility in Mexico’s outsourcing industries seem artificially high. To avoid this problem, we limit the analysis to the period 1996-2005.

To provide a visual sense of the relative variation in industry activity in the two countries, Figure 2 plots the production-worker employment for the four core outsourcing industries over the sample period. To remove seasonal fluctuations and time trends, each series is seasonally adjusted and HP filtered (with smoothing parameter 14400). In each industry, employment in Mexico is substantially more volatile than in the United States. Table 1, which shows the ratio of the standard deviations for the production worker employment in Mexican and U.S. industries, reinforces this perception. In each industry, the standard deviation of Mexican employment is greater than in the United States, with the Mexico-U.S. ratio averaging 2.2 over the four industries.

One simple reason that employment in Mexican industries could be more volatile than their U.S. counterparts is because at an aggregate level the Mexican economy is also more volatile than the U.S. economy. That reasoning does not apply for employment, however, since as shown in Table 1, employment of production workers in all manufacturing in Mexico has a lower monthly standard deviation than in the United States. This may reflect the finding in Botero et al. (2004) that Mexico has more restrictive laws regarding employment security (see note 3).

A second reason for volatility to be higher in the Mexican outsourcing industries is that they are smaller than the U.S. industries, so with idiosyncratic shocks across plants, U.S. employment

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7 Data for Mexico’s maquiladora industries go back to 1990. However, data on Mexico’s overall manufacturing sector is only available from 1993 forward, owing to a change in the construction of the series in that year.

8 This conjecture does hold for the wage bill, but the extra volatility found in the wage bill for Mexican outsourcing industries exceeds that found in Mexican manufacturing. The ratio of the wage bill in Mexico to that in the U.S. is 2.0 for outsourcing industries and 1.5 for manufacturing overall. In our working paper (Bergin, Feenstra and Hanson, 2007), we use a difference-in-difference test to confirm that the volatility of the wage bill in the Mexican maquiladoras is significantly greater than the volatility in the U.S. outsourcing industry, even after controlling for the extra volatility found in overall Mexican manufacturing.
may be smoothing out shocks. To investigate the size differences between the Mexican and U.S. industries, Table 2 reports employment in each industry, showing that in two of the four industries the U.S. is indeed much larger. We can deal with these size disparities by reducing the geographic coverage of the U.S. series. The vast majority of maquiladoras in Mexico are located in Mexican border cities and many are linked to production operations on the U.S. side of the border (Feenstra, Hanson, and Swenson, 2000). This makes U.S. border states a natural geographic unit to which to compare Mexican outsourcing industries.

In Tables 2 and 3, we compare Mexican industries to their counterparts in California and Texas, the two U.S. border states for which industry data are available. At the state level, the only series available for three-digit industries is total employment, including production and nonproduction workers. Table 2 shows that employment in outsourcing industries in California and Texas is similar in scale to Mexican industries. Table 3 shows that standard deviations and their ratios based on state employment data are broadly similar to those obtained for national data: the four outsourcing industries are at least 75% more volatile in Mexico than in California, and 66% more volatile than in Texas, whereas overall Mexican manufacturing employment is less volatile than in either state. We conclude that regardless of the comparison case, the maquiladoras have substantially greater volatility than their corresponding industries in the U.S., or than overall manufacturing.

C. Volatility in the Extensive Margin

The theoretical model we develop in the following section will imply that changes in employment by outsourcing industries are driven in part by adjustment at the extensive margin. The mechanism we have in mind is one in which aggregate shocks alter Mexican wages relative to U.S. wages, inducing an endogenous shift in the extent of production activities U.S. firms outsource to

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9 The small number of plants in three-digit manufacturing industries in Arizona and New Mexico makes industry-level data for these states subject to disclosure restrictions.
Mexico. If such a mechanism is operative, we should see considerable entry and exit among the assembly plants in Mexico that produce intermediate goods and services for U.S. industry. There is abundant anecdotal evidence of such plant turnover. During the U.S. economic expansion of the 1990s, the number of maquiladoras in Mexico grew from 1,600 to 3,700. Delphi, a large U.S. manufacturer of auto parts, expanded its operations in Mexico to include six assembly plants. As U.S. economic growth slowed in 2001 and 2002, over 700 Mexican maquiladoras closed shop. Delphi shut down two of its Mexican maquiladoras, leaving the other four in operation. Similar expansion and contraction in Mexican assembly plants is evident among firms that produce TV sets, cell phones, computer printers, and other goods.

More formal evidence of adjustment at the extensive margin comes from examining the contribution of plant entry and exit to changes in maquiladora employment. We use data on the number of plants operating each month in the maquiladora sector and the average employment per plant. Research on exporting plants in other countries has found that the sunk costs of becoming an exporter are quite high, with much smaller continuation costs of exporting. For example, Das, Roberts and Tybout (2001) estimate that the sunk costs of becoming an exporter for Colombian plants are on the order of $1 million, or between 20% and 40% of the annual export value (with continuation costs around 1%). Those values suggest that movement in and out of exporting status would operate in a lumpy fashion, and only in response to large shocks. Our data on the maquiladora plants differs from other country samples in that all of these plants are involved in outsourcing activities to the United States, so plant entry and exit is the same as the exporting decision. As we now show, the variation in the number of plants is quite large.

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The monthly standard deviation in the number of plants (logged, deseasonalized and HP filtered) ranges between 2.1% for electrical machinery to 5.2% for apparel, with an average of 3.1% over the four industries. That variation is about two-thirds as much as the monthly standard deviation in employment, and there is a positive correlation between the number of plants and employment. We can evaluate the relationship between aggregate employment and the number of plants more formally along the lines of Eaton, Kortum and Kramarz (2004), using the identity:

\[ N_{it} \times \frac{E_{it}}{N_{it}} = \frac{E_{it}}{E_t} \times E_t , \tag{1} \]

where \( E_{it} \) is employment in industry \( i \) at time \( t \), \( N_{it} \) is the number of plants in industry \( i \) at \( t \), and \( E_t \) is aggregate employment in Mexico at \( t \). From this identity we specify two regression equations:

\[ \ln N_{it} = \alpha_0 + \alpha_1 \ln \frac{E_{it}}{E_t} + \alpha_2 E_t + \varepsilon_{it} \tag{2} \]

and

\[ \ln \frac{E_{it}}{N_{it}} = \beta_0 + \beta_1 \ln \frac{E_{it}}{E_t} + \beta_2 E_t - \varepsilon_{it} \tag{3} \]

By the logic of least squares, \( \alpha_0 + \beta_0 = 0 \), \( \alpha_1 + \beta_1 = 1 \) and \( \alpha_2 + \beta_2 = 1 \). The relative magnitude of the coefficients identifies how aggregate shocks affect the number of plants (the extensive margin) and employment per plant (the intensive margin).

Table 4 reports results for this decomposition exercise. All series have been deseasonalized and HP filtered and data are pooled across the four outsourcing industries. Controls for industry fixed effects are included in the estimation. In column (1), the estimate for \( \alpha_1 \) is 0.38 and for \( \alpha_2 \) is 0.49, where both are statistically significant. In response to an increase in the share of aggregate employment in an outsourcing industry (holding aggregate employment constant) over one-third of adjustment in industry employment occurs at the extensive margin, and in response to an increase in aggregate employment (holding the industry employment share constant) nearly one-half of adjustment in industry employment occurs at the extensive margin. It thus appears that plant entry and exit is an
important mechanism through which the maquiladora industry adjusts to aggregate shocks.

A second way to measure the extensive margin, which will be close to our theoretical model, is by the number of products that Mexico exports to the U.S. We use the Harmonized System (HS) import data for the U.S., at a monthly frequency, and focus on the three largest land border crossings: Laredo, El Paso and San Diego. In Table 5, we summarize the average number of HS 10-digit products crossing at each location per month in the four industries. For example, there are 300 or more types of apparel items being imported at Laredo and San Diego each month, and a smaller number of items in other outsourcing industries. We also report the mean number of months that HS products in each industry are exported each year, ranging from 5.8 months for apparel items in El Paso, to 9.0 months for transport equipment in Laredo.

These summary statistics show that there are many ‘zeros’ in the data, i.e. many instances where an HS product is not exported some month. That fact is also illustrated by Figure 3, where we graph the log number of HS products per month, after deseasonalizing and HP filtering. The standard deviation of these series (reported in Table 5), averaged across industries in 3.6%. Some of the fluctuation in the number of HS products will reflect products that are exported at irregular intervals during the year. But we also expect that some fluctuation is systematic: visually, there appears to be a fall in the number of HS products in 2002, at the time when employment in both Mexico and the U.S. fell substantially (see Figure 2). That impression is confirmed by the correlations between the number of HS products and U.S. manufacturing employment reported in Table 5, which are nearly all positive and often exceed 0.25.

Beyond confirming that the range of products exported from Mexico is procyclical with U.S. employment, we wish to determine the degree to which there is a systematic component to its fluctuation.13 To this end, we specify the same relationship shown by (1), where \( N_{it} \) now denotes the

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13 Alessandria, Kaboshki and Midrigan (2007) argue that large devaluations leads to systematic changes in the range of goods exported, also using the monthly 10-digit HS data for the United States.
number of HS products exports from Mexico in industry \( i \), month \( t \), and a particular border crossing, \( E_{it} \) is the value of exports for that industry \( i \) and border crossing in month \( t \), and \( E_t \) is total exports at that border crossing in month \( t \). Then the regressions (2) and (3) follow as before, where \( (E_{it}/N_{it}) \) is the average value of Mexican exports per HS product. The results from these regressions are shown in columns (3) and (4) of Table 4. All series have been deseasonalized and HP filtered, and then are pooled across the four outsourcing industries, with controls included for industry fixed effects. In column (3), the estimate for \( \alpha_1 \) is 0.07, which is statistically significant, and for \( \alpha_2 \) is 0.10, which is also significant. The first coefficient shows that the number of HS products in each industry responds to a shift in Mexican exports towards that industry. In that sense, the number of products provides an avenue of adjustment, but by an amount smaller than we found for the entry and exist of plants in Mexico. For an increase in overall exports (the second coefficient), the number of HS products in the industry also responds by a modest amount.

A problem with using the number of HS products as a measure of adjustment is that the HS classification is quite arbitrary: the classifications are determined more by the need to distinguish tariffs for various items than by any consideration of the value of exports for each product, which might be very small or very large. To address that problem, we consider instead the ‘extensive margin of exports,’ as used by Hummels and Klenow (2005). Let \( J_{ikt} \) denote the set of 10-digit Harmonized System exports from Mexico to the U.S. in industry \( i \), at border crossing \( k \), and month \( t \). The extensive margin of exports is a sales-weighted count of the number of products in the set \( J_{ikt} \), relative to the number in a “common” set \( J_{ik} \), which we define as the products exported every month within the year. Then the extensive margin is defined as:14

\[ EM = \frac{\sum_{i} n_{ikt} \times \log \left( \frac{J_{ikt}}{J_{ik}} \right)}{\sum_{i} n_{ikt}} \]

14 From Feenstra (1994), this formula is the correct way to count the number of products for a CES consumer, in that the welfare gain to the consumer from having products in \( J_{ikt} \) that are not in \( J_{ik} \) is proportional to the log of \( EM \), with the factor of proportionality depending on the elasticity of substitution between products. Hummels and Klenow (2005) were the first to refer to this formula as the extensive margin of exports.
where \( \bar{p}_{ijk} x_{ijk} \) is the value of exports for product \( j \) in industry \( i \), summed across locations other than border crossing \( k \), and averaged over months within each year. By measuring the value of exports at border crossing other than \( k \), and averaging over time, we avoid endogeneity between total monthly exports at each border crossing and the values appearing in (4).

To determine the systematic component of the fluctuations in the extensive margin, we follow Hummels and Klenow (2005) in defining the intensive margin of exports \( IM_{ikt} \) by:

\[
EM_{ikt} \times IM_{ikt} = \sum_{j \in J_{ik}} p_{ijk} x_{ijk},
\]

where the right-hand side is the total exports in industry \( i \), border crossing \( k \) and month \( t \). Let us rewrite total exports as \( Share_{ikt} \times Export_{kt} \), where \( Share_{ikt} \) is the share of industry \( i \) in the total exports \( Export_{kt} \) at that border crossing. Then we have the identity:

\[
EM_{ikt} \times IM_{ikt} = Share_{ikt} \times Export_{kt}.
\]

From this identity we specify the two regression equations:

\[
\ln EM_{ikt} = \alpha_0 + \alpha_1 \ln Share_{ikt} + \alpha_2 Export_{kt} + \epsilon_{ikt},
\]

and

\[
\ln IM_{ikt} = \beta_0 + \beta_1 \ln Share_{ikt} + \beta_2 Export_{kt} - \epsilon_{ikt}.
\]

Again, by the logic of least squares, \( \alpha_0 + \beta_0 = 0 \), \( \alpha_1 + \beta_1 = 1 \) and \( \alpha_2 + \beta_2 = 1 \).

The results from these regressions are in the last two columns of Table 4. The coefficient \( \alpha_1 \), which indicates how the extensive margin responds to the export of industry \( i \), rises from 0.07 to 0.13. It remains highly significant, indicating that the extensive margin responds systematically to changes in total industry exports. The coefficient \( \alpha_2 \) falls slightly from 0.10 to 0.09 and remains significant, indicating that the extensive margin also responds systematically to changes to overall
exports at that border crossing. We conclude that the changes in the extensive margin – measured by either the number of maquiladora plants in Mexico or the range of product crossing the border – are margins of adjustment that respond at a monthly frequency. We now turn to a model that emphasizes these margins of adjustment.

III. Theoretical Model

A. Pricing and Product Variety in the Outsourcing Sector

In this section we develop a theoretical model of outsourcing that can broadly replicate the empirical finding documented above, and that helps us understand how outsourcing generates amplified volatility in a country like Mexico. Consider a model of two countries, labeled home and foreign. The outsourcing relationship in the model is characterized by the home country outsourcing to the foreign country, so that home may be thought of as representing the U.S. and foreign as representing Mexico. We will scale the quantity variables coming from the foreign country by its relative size: if the share $n$ of the world population resides at home and $1-n$ in the foreign country, then we scale foreign quantities by $(1-n)/n$. Foreign variables will be denoted by an asterisk *.

Each country has two sectors. The first is a standard homogeneous good whose production is specific to that country; this country-specific sector will be subscripted by $H$ for the home produced good and $F$ for the foreign produced good, or by $X$ when referring to a composite of the home and foreign homogenous goods in the consumption bundle. The second sector consists of differentiated products that are multinational, subscripted by $M$, in that they can be produced using factors in either country. This sector represents the aggregate of the four industries listed in the empirical section above, and it sometimes will be referred to as the outsourcing sector. There is a continuum of products in this sector indexed by $z \in [0,1]$, and for each $z$, there is free entry of firms who then produce $N(z)$ differentiated varieties of good $z$. The model follows Romalis (2004) in combining a continuum of products $z$ in the $M$ sector along with multiple varieties $N(z)$ of each product.
Production in the outsourcing sector involves a fixed cost activity as well as a variable cost activity. The fixed cost, $B$, represents headquarters and R&D services, and is paid every period. It is assumed here to be uniform across goods and takes place in the home country, due to the assumption that it is sufficiently more productive in these activities. The variable cost activity has a unit labor input requirement that differs by product, and follows the function $a_M(z) = \exp(az + b_t)$ in the home country. The foreign country has a corresponding function, and the relative unit cost function between the two countries will be specified as,

$$A(z) = \frac{a_M(z)}{a_M^*(z)} = \frac{\exp(az + b_t)}{\exp(a^*z + b^*_t)} = \exp(a_d z + b_{dt}),$$

where $a_d \equiv a - a^*$ and $b_{dt} \equiv b_t - b^*_t$. By ordering of the products $z$, we assume that $A(z)$ is decreasing in $z$, i.e. products are arranged by increasing order of home comparative advantage, so that $a_d < 0$. It follows that those outsourcing activities below some cutoff $z'_c$ will be produced in the foreign country, while those activities above $z'_c$ will be done at home. The cutoff activity $z'_c$ is determined by the equality of unit-labor costs in the two countries, or given the wages $W_t$ and $W_t^*$, by:

$$A(z'_c) = \frac{W_t^*}{W_t}.$$  \hspace{1cm} (8)

Overall home-country demand in this multinational sector is specified as,

$$\ln D_{Mt} = \int_0^1 \ln d_{Mt}(z)dz,$$  \hspace{1cm} (9)

where $d_{Mt}(z)$ is the demand for a product $z$. Each of these product market $z$ is assumed to be monopolistically competitive, with demand for a product specified as a CES aggregate over individual varieties $i$,

$$d_{Mt}(z) = \left[ \int_0^{N_t(z)} d_{Mt}(z,i)^{\frac{\sigma-1}{\sigma}} di \right]^{\frac{\sigma}{\sigma-1}}.$$
The composite multinational good $D_{Mt}$ will serve as the numeraire.

To determine the number of firms we make use of zero profits for each product $z$. Since the price of the multinational good is taken as numeraire, the revenue earned from home plus foreign sales of each variety in a symmetric equilibrium equals $[D_{Mt} + D_{Mt}^* (1 - n) / n] / N_t(z)$. To obtain profits, equal to revenue minus variable costs, we divide by the elasticity of demand, $\sigma$. Setting this equal to fixed costs $BW_t$, we then solve for the number of products. Under our assumptions that the fixed cost and weight in demand are uniform across varieties, the number of firms likewise is uniform across varieties, so $N_t$ does not vary with $z$:

$$N_t = \frac{D_{Mt} + D_{Mt}^* \left( \frac{1 - n}{n} \right)}{\sigma BW_t}.$$  (10)

The overall labor demand in the multinational sector at home includes labor used for the fixed cost activity, $L_{Bt}$, as well as the variable cost activity, $L_{Mt}$, for activities not outsourced ($z_i > z_i^*$); labor demand abroad includes just variable cost activity for outsourced activities ($z_i < z_i^*$), so that:

$$L_{Bt} = N_t B = \left( \frac{D_{Mt} + D_{Mt}^* \left( \frac{1 - n}{n} \right)}{\sigma W_t} \right)$$  (11)

$$L_{Mt} = \left( \frac{D_{Mt} + D_{Mt}^* \left( \frac{1 - n}{n} \right)}{\sigma W_t} \right) (\sigma - 1) \left( 1 - z_i^* \right)$$  (12)

$$\left( \frac{1 - n}{n} \right) L_{Mt}^* = \left( \frac{D_{Mt} + D_{Mt}^* \left( \frac{1 - n}{n} \right)}{\sigma W_t} \right) (\sigma - 1) z_i^*$$  (13)
B. Production in the Rest of the Economy

The remainder of the model follows a standard open macroeconomy specification. The country-specific sector in the home country is a perfectly competitive market for the homogeneous traded good with production function:

\[ Y_{ht} = \frac{L_{ht}}{a_{ht}}, \quad (14) \]

where \( L_{ht} \) is labor in the home country-specific sector and \( a_{ht} \) is its unit labor input requirement. Profit maximization by producers in this sector implies,

\[ W_t = \frac{P_{ht}}{a_{ht}}, \quad (15) \]

where \( P_{ht} \) is the relative price of the home domestic good in terms of the multinational good numeraire. Analogous conditions apply to the foreign country’s homogeneous good.

C. Households

Household preferences in the home country are represented by an instantaneous utility function of consumption \( (C_t) \), which is a composite of goods in the three sectors, and overall labor \( (L_t) \):

\[
U_t = \frac{1}{1 - \phi} C_t^{1 - \phi} - \frac{1}{1 + \mu} L_t^{1 + \mu}
\]

where

\[
C_t = \left( 1 - \alpha \right)^{\frac{1}{\xi}} C_{x_t}^{\frac{x_t - 1}{x_t}} + \alpha^{\frac{1}{\xi}} \left( C_{Xt}^{\frac{x_t - 1}{x_t}} \right)^{\frac{x_t}{x_t - 1}}
\]

\[
C_{x_t} = \left( \theta^n C_{X_t}^{\frac{q - 1}{q}} + (1 - \theta)^n C_{P_t}^{\frac{q - 1}{q}} \right)^{\frac{q}{q - 1}}
\]

In each period \( t \), the economy experiences one of finitely many events, \( s_t \). We denote by \( s' = (s_0, ..., s_t) \) the history of events up through and including period \( t \). All variables referred to so far are implicitly functions of the states of nature in period \( t \) (e.g., \( Y_{ht} = Y_{ht}(s') \)). We assume complete
contingent asset markets, as discussed more fully in Chari, et al. (2002). In period \( t \), consumers in either country purchase state-contingent assets in units of the numeraire good, denoted by 
\[ B_{s,t} = B(s^{t+1}) \], which bear a return of exactly one unit of the numeraire good in period \( t+1 \) if state \( s^{t+1} \) occurs. They purchase these assets at the prices \( V(s^{t+1} | s^t) \), which denotes the price of one unit of the numeraire good at \( s^{t+1} \) in units of the numeraire good at state \( s^t \). The budget constraint facing the home household in period \( t \) is therefore:

\[
P_t C_t \sum_{s^t} V(s^{t+1} | s^t) B(s^{t+1}) = W_t L_t + B_t - G_t,
\]

where \( P_t \) is the price index of the home country consumption basket in terms of the numeraire good, and \( G_t \) is government spending. Analogous conditions apply for the foreign country.

Labor is mobile between sectors within a country, and between fixed and variable cost activities within the home outsourcing sector, but there is no labor mobility between countries, meaning each country has a single but distinct equilibrium wage rate.

Households maximize the expected discounted sum of current and future instantaneous utility defined above, using the discount factor \( \beta \), subject to the budget constraint.\(^{15} \) The first order conditions for this problem imply the following. Risk-sharing equates the marginal utilities of consumption up to a constant of proportionality,

\[
\frac{P_t C_t}{P_t C_t^{s^t}} = \omega,
\]

where \( \omega \) is a constant indicating the relative per-capita wealth of the home country in the initial asset allocation. Relative demands for the domestic and multinational goods are:

\[
\frac{C_{x_t}}{C_{x_t}} = \frac{\alpha}{1-\alpha} \left( \frac{P_{x_t}}{P_{x_t}} \right)^{-\alpha}
\]

\(^{15} \) As the discount factor \( \beta \) is assumed identical across countries, it cancels out of the risk sharing condition (16), and does not appear in the other equations of the model.
and,
\[
\frac{C_{It}}{C_{Ft}} = \left( \frac{\theta}{1-\theta} \right) \left( \frac{P_{It}}{P_{Ft}} \right)^{-\eta}
\] (18)

where,
\[
P_{Xt} = \left( \theta P_{It}^{1-\eta} + (1-\theta) P_{Ft}^{1-\eta} \right)^{1 \over 1-\eta}.
\]

Note that the law of one price holds, so the relative prices \(P_{It}\) and \(P_{Ft}\) apply to the goods markets in both countries. The labor supply is,
\[
L_t^\mu = \frac{W_t}{P_t} C_t^{-\phi}.
\] (19)

Corresponding conditions apply for the foreign country.

D. Market Clearing and Equilibrium

Exogenous government consumption, \(G_t\), is introduced to capture shocks to demand. It is assumed that the home government allocates demand between homogeneous domestic and differentiated multinational good in the same fashion as consumers, except that among homogenous goods it consumes only the home and not the foreign good. This implies \(G_{Mt} = \alpha \left( \frac{P_{Mt}}{P} \right)^{-\chi} G_t\),
\[
G_{Mt} = \left( 1-\alpha \right) \left( \frac{P_{Mt}}{P} \right)^{-\chi} G_t \quad \text{and} \quad G_{Ht} = G_{Mt} \left( \frac{P_{Mt}}{P_{It}} \right).
\]
This holds analogously for the foreign government’s demand. We then define the home country demand for the multinational sector noted in the preceding section as \(D_{Mt} = C_{Mt} + G_{Mt}\). The market clearing condition for the homogeneous good of home country is:
\[
C_{It} + \left( \frac{1-n}{n} \right) C_{It} + G_{It} = Y_{It}.
\] (20)

Equilibrium in the labor market requires that overall labor supply equal the sum of labor demands across sectors:
\[
L_t = L_{It} + L_{Mt} + L_{Ht},
\] (21)
The general equilibrium is a sequence of 19 endogenous variables: \( L_t, L^*_t, L_{Ht}, L^*_{HT} \), \( L_{Mt}, L^*_{Mt}, W_t, W^*_t, N_t, C_{Ht}, C^*_{Ht}, C_{Ft}, C^*_{Ft}, C^*_{Mt}, P_{Ht}, P_{Ft} \), and \( z^* \), which are determined by the labor-supply condition (19), relative demand for the multinational and home country-specific goods (17) and (18), labor demand for the country-specific sector (14) and the multinational sector (12), the market clearing condition for the country-specific sector (20), the market clearing condition for labor (21), and the foreign counterparts for each of these. In addition, there is the marginal outsourcing condition (8), the risk sharing condition in (16), the zero profit condition (10), labor demand for fixed cost labor (11), and the normalization of the price of the numeraire good \( D_{Mt} \) (as described in the Appendix).

E. Shocks

The model will include shocks both to demand and supply, entering through the additive demand terms (\( G_t \) and \( G^*_t \)) and the unit labor requirement terms (\( a_{Ht} \) and \( a^*_{Ft} \)). Both types of shocks are specified as first order autoregressions in log deviations from their respective means:

\[
\begin{bmatrix}
\log(a_{Ht}) - \log(\bar{a}_{Ht}) \\
\log(a^*_{Ft}) - \log(\bar{a}^*_{Ft})
\end{bmatrix} = \rho_a \begin{bmatrix}
\log(a_{Ht-1}) - \log(\bar{a}_{Ht-1}) \\
\log(a^*_{Ft-1}) - \log(\bar{a}^*_{Ft-1})
\end{bmatrix} + \begin{bmatrix}
\varepsilon_{aHt} \\
\varepsilon_{aFt}
\end{bmatrix}, \text{ where } \begin{bmatrix}
\varepsilon_{aHt} \\
\varepsilon_{aFt}
\end{bmatrix} \sim \mathcal{N} \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \sigma_a \right)
\]

and:

\[
\begin{bmatrix}
\log(G_t) - \log(\bar{G}) \\
\log(G^*_t) - \log(\bar{G}^*)
\end{bmatrix} = \rho_G \begin{bmatrix}
\log(G_{t-1}) - \log(\bar{G}) \\
\log(G^*_{t-1}) - \log(\bar{G}^*)
\end{bmatrix} + \begin{bmatrix}
\varepsilon_{Gt} \\
\varepsilon_{G^*t}
\end{bmatrix}, \text{ where } \begin{bmatrix}
\varepsilon_{Gt} \\
\varepsilon_{G^*t}
\end{bmatrix} \sim \mathcal{N} \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \sigma_G \right).
\]

We do not compute a separate set of Solow residuals for the outsourcing sector of each country, as this might end up replicating the higher volatility in the Mexican outsourcing sector in a purely mechanical way. Instead, we consider two cases of how a country’s domestic productivity shocks affect the outsourcing sector. One case simply assumes that productivity moves proportionately in the domestic and outsourcing sectors within each country: \( \tilde{a}_{Mt}(\tau^*) = \tilde{a}_{Ht} \), \( \tilde{a}^*_{Mt}(\tau^*) = \tilde{a}^*_{Ft} \), where tildes
indicate log deviations from means. This specification implies that the relative cost schedule \( A(z) \) shifts in response to productivity shocks. The other case differs in that it holds the \( A(z) \) schedule fixed, assuming that productivity shocks in a country’s outsourcing sector are perfectly transmitted to the other country’s outsourcing sector: \( \bar{a}_{Mt}(\bar{z}^*) = \bar{a}_{Mt}^*(\bar{z}^*) = \bar{a}_{Ht} + \bar{a}_{Fr}^* \). This specification is similar to that assumed in Burstein et al. (2005), and is also consistent with Berman, Bound and Machin (1998) who argue that technology shocks spread quickly to other countries. We take this second specification as our benchmark, although we shall report results for the first case as well.

IV. Model Results

A. Analytical Results

Some important intuition can be gained into relative volatilities across countries by using the labor demand conditions for the variable cost activity in (12) and (13). Denote the world demand for the multinational good as \( D^w_{Mt} = [D_{Mt} + D^*_w(1 - n)/n] \), and replace \( W_t \) in (23) by \( W_t^*/A(z^*_t) \) using (8), to obtain:

\[
L_{Mt} = \left( \frac{D^w_{Mt}}{W_t^*} \right) (1 - z^*_t) A(z^*_t) \left( \frac{\sigma - 1}{\sigma} \right) \quad (23)
\]

\[
L^*_{Mt} = \left( \frac{D^w_{Mt}}{W_t^*} \right) z^*_t \left( \frac{n}{1 - n} \right) \left( \frac{\sigma - 1}{\sigma} \right) \quad (24)
\]

We focus on the variable cost activity in these equations, as our data apply to production workers in the two countries.\(^\text{16}\) Recalling \( A(z_t') = \exp(a_d z_t' + b_{dt}) \) with \( a_d < 0 \), let us further focus on demand shocks by treating \( b_{dt} \) as constant. Then taking natural logs of (23)-(24) and using a first-order approximation for \( \ln z_t' \) and for \( \ln(1 - z_t') \), we can compute the variances:

\(^\text{16}\) The choice to exclude fixed cost labor from the U.S. total potentially works against our model’s ability to generate the desired lower volatility in the U.S. But this measure is the appropriate comparison to the data, which measure employment of production workers. Simulation results are nearly identical if we include fixed cost labor, as (12)-(14) indicate that \( W_t L_{Mt} \) is proportional to \( W_t (L_{Mt} + L_{Fr}) \), including fixed costs at home, for fixed \( z' \).
Thus, the variance of employment in the multinational sector depends on the variance of world demand relative to the foreign wage; on the variance of the outsourcing margin $z_i$ measured relative to an ‘adjusted’ size of the outsourcing sector, $\{(1/(1-z'))-a_d\}$ at home and $1/(1-z')$ abroad; and also on the covariance between $z_i$ and world demand.

The covariance term in (25)-(26) will generally be positive for home demand shocks, since an increase in home demand raises the outsourcing margin $z_i$ but also lowers the foreign wage $W_{Mt}^*$. Notice that the covariance enters with a positive sign in (26) but a negative sign in (25): as more outsourcing activities are shifted abroad, that will tend to offset the variance in home employment in the outsourcing industry, but enhance it abroad, because the procyclical nature of $z_i$ will amplify the impact abroad of the demand shocks. Therefore, a positive covariance between world demand for the multinational good and the outsourcing margin will shift variance towards the foreign country. This situation is clearly relevant for the U.S./Mexico case we study, as the U.S. is the primary demander for outsourced goods both in the data and in our simulation due to its larger population and wealth.

The volatility in employment also depends on the size of the outsourcing sector in the two countries, adjusted by $a_d < 0$ which determines the slope of the $A(z)$ schedule. Provided that

\[
\{1/(1-z')\}-a_d < 1/(z')
\]

which is easily satisfied in our calibration, then the volatility of

\[\text{var}(\ln L_{Mt}) = \text{var}\left[\ln\left(\frac{D_{Mt}^w}{W_{Mt}^*}\right)\right] + \frac{1}{1-z'} \text{var}(z_i) - 2 \left[\frac{1}{1-z'} - a_d\right] \text{cov}[z_i, \ln\left(\frac{D_{Mt}^w}{W_{Mt}^*}\right)], \quad (25)\]

and,

\[\text{var}(\ln L_{Mt}^*) = \text{var}\left[\ln\left(\frac{D_{Mt}^w}{W_{Mt}^*}\right)\right] + \frac{2}{1-z'} \text{cov}[z_i, \ln\left(\frac{D_{Mt}^w}{W_{Mt}^*}\right)]. \quad (26)\]

\[17 \text{ With } h_{dt} \text{ constant, it is readily shown from (A3) in the Appendix that any increase in the home wage due to demand shocks must be matched by a decrease in the foreign wage, since both are measured relative to the price of the multinational good which is the numeraire. Therefore, home demand shocks raise } W_{t} \text{ and lower } W_{t}^*. \]
employment in the foreign country will magnified by the variance in \( z'_f \). This occurs simply because the shift of a given amount of employment from home to foreign will have a greater percentage impact abroad when the foreign country has a smaller outsourcing sector. So for this reason, also, the foreign country will experience greater volatility.

Our model also includes supply shocks in the two countries, but their impact is sensitive to how the shocks are specified. In our benchmark case we hold fixed the \( A(z) \) schedule. In that case, supply shocks in the homogeneous goods turn out to have minimal impact on the outsourcing margin \( z' \), relative wages, or other endogenous variables. This finding follows from a little-known result in the international macro literature. Consider our model without the outsourcing sector, so that it is a two-good Ricardian model with complete asset markets, and assume Cobb-Douglas preferences over the homogeneous goods from each country. Then it turns out that the technology shocks have no impact at all on relative wages. The reason is that a positive productivity shock to, say, the home good, will reduce its price on international markets but increase its demand by the same percentage amount, so the implied relative labor demand across countries is unchanged. Relative labor supply is also unchanged due to complete asset markets, so the equilibrium relative wage does not change.\(^{18}\)

Adding the outsourcing sector but with a fixed \( A(z) \) schedule leads to nearly the same result. However, when we allow the \( A(z) \) schedule to respond to technology shocks, then a rise in home productivity shifts the \( A(z) \) schedule down compared to a given relative wage, inducing a fall in outsourcing. This “in-sourcing” behavior implies a negative correlation between employment in the

\(^{18}\) The result that neutral technology shocks have no impact on the relative wage in a two-sector model with large countries and Cobb-Douglas preferences is noted by Krugman (2000), in a model with fixed labor endowments. Using labor choice and complete asset markets, Devereux and Engel (2001) obtain a factor-price equalization result under the assumption that labor enters the utility function linearly. Our assumptions are weaker, since labor enters utility with a power. Combining (16) and (19) we readily obtain \( (L_r / L_f)^\mu = (W_r / W_f^*) \), which is the relative labor supply schedule. Relative labor demand in the two-sector Ricardian economy is obtained by combining (14) with Cobb-Douglas preferences, \( L_{H} / L_{F} = a_{H}Y_{H} / a_{F}Y_{F} = (a_{H} / a_{F})(p_{F} / p_{H}) \). Then substituting for prices from (15) we have, \( L_{H} / L_{F} = (a_{H} / a_{F})(a_{F}W_{F} / a_{H}W_{H}) = (W_{F} / W_{H}) \). This proves that both relative labor demand and supply do not depend on the technology shocks, so neither does the equilibrium relative wage.
home and foreign outsourcing activity, and is counterfactual given the positive correlations observed in the data. This is one benefit of using our benchmark specification of productivity shocks, where the $A(z)$ schedule is held fixed by assuming transmission of productivity across countries within the outsourcing sector.

**B. Calibration of Numerical Experiment**

We calibrate the model to monthly data for the U.S. and Mexico. Parameter choices are summarized in Table 6. We have already argued that the magnitude of $z'$ is important. Taking the ratio of (12) and (13) indicates that $z'$ can be calibrated to match the Mexican share of production labor earnings in the outsourcing sector.

$$\frac{(1-n)W^*_tL'_{Me}}{nW_tL_{Me}} = \frac{z_i}{1-z_i}$$

(Averaging over our data for earnings in the four outsourcing industries at the end of our sample indicates a share of 0.06.

The U.S. unit cost schedule in the outsourcing sector is characterized by the two parameters, $a$ and the mean of $b$. First, we calibrate the level of this schedule at the steady state margin to be the same as for the overall U.S. manufacturing sector, which is normalized to unity (that is, $a_M(z') = \exp(a z' + b) = \bar{a}_H = 1$). Second, we replicate the observation in Bernard et al. (2003) that the standard deviation of log U.S. plant sales is 1.67. This is sufficient to pin down both schedule parameters.19

Similarly, two pieces of information are needed to pin down the two parameters $a_d$ and $b_d$ in the cross-country relative unit cost schedule, $A(z_t) = \exp(a_d z_t + b d_t)$. First, we calibrate the level of this schedule at the steady state outsourcing margin to our data. The average weekly rate of payment

19 In particular, given that the schedule is defined over the unit interval, and if we identify firm size with sales, which depends on price and the elasticity $\sigma$, the constant value $a$ can be computed as $-1.67(12)^{0.5}/\sigma$. Then $b_t$ varies with overall productivity shocks as $b_t = \log(a_{\sigma}) - a\bar{z}$.
to workers for the four outsourcing industries in the data set is 8 times higher in the U.S. than in Mexico, implying that \( A(\zeta') = \frac{1}{8} \) (and hence \( \frac{\theta_d}{d} = \ln(0.125) - a_d \zeta' \)). Secondly, we calibrate the slope of the schedule at the steady state margin, by using estimates from Eaton and Kortum (2002) of the variation within the distribution of productivity across a continuum of goods.\(^\text{20}\) They show that their distribution implies a relative distribution of unit labor input requirements across two countries of \( A(z) = \left( \frac{T_1}{T_2} \right)^{1/\theta} \left[ \frac{(1 - z)}{z} \right]^{1/\theta} \), where \( \theta \) is the parameter characterizing variation in the distribution within the continuum of goods, and \( T_1 \) and \( T_2 \) characterize the absolute technology advantage of the countries over the whole continuum. This is analogous to the relative cost schedule of Dornbusch et al. (1976), and thereby analogous to our own \( A(z) \) schedule. Given their preferred estimate of \( \theta \) at 8.28, and since the ratio \( T_1/T_2 \) is pinned down by our calibration above that \( A(\zeta') = \frac{1}{8} \), this is sufficient information to pin down the desired slope as \( A'(\zeta') = -0.2677 \), and hence \( a_d = A'(\zeta') / A(\zeta') = -2.14 \).

The fixed cost, \( B \), enters the model only jointly as a product with \( N \), the number of firms, where \( NB \) represents the total fixed costs paid by an industry in labor units. Domowitz, et al. (1988) report estimates of the size of fixed costs for a range of U.S. industries as a share of industry output. Three of these correspond to the industries we use (apparel, electronic equipment, and transportation equipment), and the average among these indicates a share of 0.11. Equations (10) and (12) imply that this share can be written as \( \frac{WBN}{WL_M} = 1/[1 + (\sigma - 1)(1 - \zeta')] \). Given calibration of \( \zeta' \) above, matching the fixed cost share necessarily pins down the calibration of \( \sigma \) at 9.5.

The relative wealth parameter \( \omega \) is chosen to imply a ratio of per-capita consumption of the U.S. to Mexico in steady state that matches the data, which the Penn World tables indicate is 4.3 in 2000. The parameter \( n \) is calibrated to reflect the 74 percent share of the U.S. in the combined

\(^\text{20}\) Their estimates are based upon bilateral import shares and price differences for 50 manufactured products across 19 countries in 1990.
population of the two countries. Calibrations of standard preference parameters are taken from the business cycle literature. The labor supply elasticity is set at unity, $\mu = 1$. The curvature parameter is set at $\phi = 2$. The elasticity of substitution between home and foreign goods is calibrated at the common value of unity, $\eta = 1$.\(^{21}\) We assume a higher elasticity, $\chi = 2$, between the homogeneous and multinational good.

The remaining preference parameters are calibrated to reflect the relationship between U.S. and Mexican aggregates. The home bias parameters reflect the share of import expenditures in GDP, $\theta = 0.88$, $\theta^* = 0.71$. The four U.S. industries classified as outsourcing industries in the data set represent 24% of total U.S. manufacturing, so the outsourcing share parameter is calibrated at $\alpha = 0.24$. The relative weight on the home country in the complete asset market allocation ($\varpi$) is calibrated to replicate the ratio of U.S. to Mexican per capita consumption. The steady state levels of the additive consumption terms ($\bar{G}$ and $\bar{G}^*$) are calibrated at 15% of total demand, representing the share of government purchases in U.S. gross national expenditure during our sample range.

The mean unit labor requirement in Mexico’s domestic sector ($\overline{a}_F$) is calibrated to imply a steady state of the model consistent with the assumption above that the Mexican wage is 1/8 that of the U.S. Productivity shock parameters are estimated from a first-order autoregression on Solow residuals, computed from monthly manufacturing data from our sample range.\(^{22}\) Regarding demand shocks, since there is no monthly series available for government consumption, total government spending from IFS is used. Linear trends are removed from the log of each series before fitting it to the first-order autoregressive processes used for all shocks above. See Table 6 for exact values.

### C. Numerical Results

\(^{21}\) Bergin (2006) estimates a value near unity for this elasticity using macroeconomic data.  
\(^{22}\) We follow the convention in Glick and Rogoff (1995) of computing Solow residuals by setting the labor share at 0.6 and assuming a constant capital stock. Resulting estimates are almost identical if we assume a labor share of unity, as implied by the production function in the model above.
Simulations consist of solving the model numerically in its original nonlinear form for 120 periods of random draws of shocks. The first 20 periods are dropped, and the remaining 100 periods are HP filtered, just as were the data reported in Table 1, and used to compute moments. This process is repeated 1000 times, and we report the average of moments over the replications.

Results for the benchmark case of the model are summarized in column (2) of Table 7, which can be compared to the data for Mexico and the U.S. in column (1). Although the focus of our study is on the outsourcing sector, it is reassuring for our general calibration of shocks that the volatilities for overall manufacturing employment are in the neighborhood of what is observed in the aggregate data, including the fact that overall employment volatility is somewhat higher in the U.S. than in Mexico. Of primary interest is the fact that the calibrated model can easily generate double the volatility in the outsourcing sector of Mexico relative to the corresponding U.S. sector. The standard deviation of employment for Mexican outsourcing, 4.5%, is remarkably close to that in the data at 4.4%; the employment volatility in U.S. outsourcing is somewhat underestimated compared to the data, 1.5% and 2.0%, respectively.\textsuperscript{23}

We next evaluate the model’s implications for extensive margin movement. For a measure of the standard deviation of $z'$, we focus on the number of HS products exported from Mexico, which averaged 3.6%. A somewhat smaller number of 2.9% is obtained in the benchmark simulation. This indicates that our explanation for Mexican volatility does not rely upon an unreasonably high degree of firms switching their outsourcing decision. We interpret our data on the number of maquiladora plants as applying to either the number of plants $N_t$, or to $N_t z_t'$, which is the product of plants per good and the mass of outsourced goods. The standard deviations for $N_t$ and $N_t z_t'$ implied by the simulation encompass the standard deviation on the number of plants from our

\textsuperscript{23}The model does not replicate the volatility of the wage-based real exchange rate across countries, however. The standard deviation of log $W*/W$ in the simulation is 0.48; in data, the standard deviation of log $s w*/w$, where $s$ is the nominal exchange rate and $w$ and $w*$ are wage indexes from IFS, is 4.22 for our sample range. The failure to explain volatility of real exchange rates of various types is common to our class of models that exclude nominal shocks and nominal rigidities. This property is not a result of endogenous outsourcing dampening relative wage volatility, as the volatility of the relative wage is likewise small, 0.87, when we suspend the $A(z)$ schedule and hold $z'$ fixed.
Regarding employment correlations, all are positive as in the data, indicating positive comovement between the outsourcing sectors of both countries. This coincides with our intuition provided earlier, for how demand shocks in particular should generate volatility in the Mexican outsourcing sector. Our model has a somewhat stronger correlation of the range of products \( z' \) with U.S. employment than obtained from the data, however.

At the bottom of Table 7 we report the theoretical covariance between \( z' \) and \( \ln(D_{Mt}^w/W_{t}^r) \). That covariance is positive in our benchmark results in column (2), reflecting the procyclical nature of the extensive margin of outsourcing. From our results in (25)-(26), that positive covariance combined with the smaller size of the outsourcing sector in the foreign country \( (\pi' = 0.06) \), both contribute to higher volatility of employment abroad.

Columns (3)-(6) indicate the results obtained when just one of the four shocks is used. These results reveal that the home demand shock is the most important driver of the amplified volatility in the Mexican outsourcing sector. Column (7) confirms the analytical claim above that our main result depends crucially on endogenous movement in the outsourcing margin. Mexican employment volatility falls to a level much closer to the U.S for outsourcing sector earnings when \( z' \) is held fixed.

The theoretical model did not attempt to represent the complex dynamics of firm entry, in which it is reasonable to think there are substantial delays between a decision to enter and actually commencing production. Nor does the theory model the sunk cost of entry, which might discourage entry in response to transitory shocks. We test whether the modeling of this feature could be quantitatively relevant for the issue at hand, by modeling it in its most extreme form. Column (8) of Table 7 reports simulation results for the case where the number of firms is held constant. The results are almost the same as the benchmark case. Since modeling the most extreme form of sluggishness in firm entry is not considered, our results are likely to be an underestimate of the true impact of this factor.

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24 If the maquiladora plants all produced a single product, then it would be appropriate to measure their number by \( N_I z' \); but if the plants all produce the full range of products, then their number should be measured by \( N_I \).
firm entry has a quantitatively small impact on the results, it would seem that modeling the intermediate cast of realistic entry dynamics is not warranted. One might also consider modeling an adjustment cost discouraging switching in the outsourcing decision, which is more likely to affect our results, given that endogenous movement in the outsourcing sector is behind the cross-country difference in volatilities. But given that the standard deviation of the outsourcing margin in our benchmark case is close to that in the data, there appears to be no empirical motivation for introducing such an adjustment cost in this model.

When the relative productivity schedule $A(z)$ is made flatter, where the slope $A'(\bar{z}^*) = -0.1$, column (9) shows that Mexican volatility rises. A flatter distribution implies greater movement in the outsourcing margin $z$.

As the analytical result identified the mean of the outsourcing margin as important for our result, column (10) reports results for the case of an equal split of goods between the U.S. and Mexico, $\bar{z}^* = 0.5$. In this case there is much less volatility in the foreign outsourcing sector than found with our benchmark calibration, though volatility there still exceeds that found in the home outsourcing sector. The reason why foreign volatility is higher even with equal-sized outsourcing industries in the two countries is that the covariance between $z^*$ and $\ln(D^W_{it} / W^*_t)$ still adds to the volatility of employment abroad, but offsets it at home, as shown by (25)-(26).

Finally, we study alternative specifications of the supply shock that are allowed to shift the $A(z)$ schedule in the outsourcing sector, as discussed previously. In column (10) this is found to further amplify the volatility in the Mexican outsourcing sector to remarkably high values. But this comes at the cost of implying strong negative correlations between the outsourcing sectors of both countries. This arises from the fact, as discussed previously, an increase in U.S. relative productivity exceeding the increase in the U.S. relative wage induces U.S. firms to “in-source” activity previously allocated to Mexico.
V. Conclusions

Global production sharing is responsible for a substantial portion of world trade and is the primary means through which many developing countries engage in international commerce. While the expansion of export assembly operations have lead to impressive rates of employment growth in China, Eastern Europe, Mexico, and elsewhere, the implications of global production sharing for the volatility of economic activity has only recently attracted scholarly attention.

In the case of Mexico, we document that the country’s outsourcing industries experience fluctuations in economic activity that are twice as volatile as the corresponding industries in the U.S. Fluctuations in outsourcing employment in Mexico come about in large part from adjustment at the extensive margin, as assembly plants enter and exit production and begin and end product lines.

To account for differences in U.S. and Mexican employment variability, we develop a theoretical model in which heterogeneous firms in a high wage country (the U.S.) are free to enter and exit outsourcing relationships with firms in a low wage country (Mexico). Shocks that change U.S.-Mexico relative wages induce U.S. firms to alter which products they outsource to Mexico. Adjustment in the outsourcing margin is the main mechanism through which U.S. shocks become amplified in Mexico. Stochastic simulations show that the model matches the empirical regularities observed in U.S. outsourcing to Mexico. For Mexico, one consequence of economic integration with the U.S. appears to be greater variability in its manufacturing employment.
References


Appendix

Labor demand at home is obtained by integrating over the fixed costs $B$ for every product $z \in [0,1]$, and the variable labor input requirement $a_{Mh}(z)y_i(z)$ for those products $z \in [z',1]$:

$$L_i = \int_0^1 BN_i dz + \int_{z_i}^1 a_{Mh}(z)y_i(z)N_i dz .$$  \hspace{1cm} (A1)

The number of varieties $N_i$ appearing in the first integral of (A1) is obtained from (10). For the second integral, we multiply the labor input requirement $a_{Mh}(z)$ by the wage $W_i$, and further multiply by the markup $\sigma/(\sigma - 1)$, to obtain the expenditure $[C_{Mh} + C_{Mh}^*(1-n)/n] / N_i$ on each variety. So the expression inside the second integral of (A1) equals $[C_{Mh} + C_{Mh}^*(1-n)/n](\sigma - 1)/\sigma W_i$, which is integrated over $z \in [z',1]$ to yield (11).

For foreign labor demand we integrate the variable labor input requirement $a_{Mh}^*(z)y_i^*(z)$ for $z \in [0,z'_t]$:

$$L_i^* = \int_0^{z_t} a_{Mh}^*(z)y_i^*(z)N_i dz .$$  \hspace{1cm} (A2)

Multiplying the labor input requirement $a_{Mh}^*(z)$ by the wage $W_i^*$, and further multiplying by the markup $\sigma/(\sigma - 1)$, we again obtain the expenditure $[C_{Mh} + C_{Mh}^*(1-n)/n] / N_i$ on each variety. So the expression inside the integral of (A2) equals $[C_{Mh} + C_{Mh}^*(1-n)/n](\sigma - 1)/\sigma W_i^*$, which is integrated over $z \in [0,z'_t]$ to yield (12).

Finally, the price index for multinational goods is:

$$\ln P_{1h} = z'_t \ln W_i^* + (1-z'_t)\ln W_i + \ln \left( \frac{\sigma}{\sigma - 1} \right) + \left( \frac{a^* - a}{2} \right)(z'_t)^2 + (b^*_t - b_t)z'_t + \frac{a}{2} + b_t ,$$  \hspace{1cm} (A3)

is set equal to unity to close the model.
Table 1. Relative Volatility in Mexico and U.S. Outsourcing Industries:
Production Worker Employment

<table>
<thead>
<tr>
<th></th>
<th>Apparel</th>
<th>Electrical Machinery</th>
<th>Computer &amp; Electronics</th>
<th>Transport Equipment</th>
<th>Average</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Standard Deviations, Employment</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \sigma(L^i) ) (Mex. Outsourcing Industry)</td>
<td>4.52</td>
<td>4.34</td>
<td>5.95</td>
<td>2.96</td>
<td>4.44</td>
</tr>
<tr>
<td>( \sigma(L^u) ) (U.S. Outsourcing Industry)</td>
<td>1.89</td>
<td>1.79</td>
<td>3.06</td>
<td>1.42</td>
<td>2.04</td>
</tr>
<tr>
<td>( \sigma(L^m) ) (Mex. Aggregate Manufacturing)</td>
<td>0.89</td>
<td>0.89</td>
<td>0.89</td>
<td>0.89</td>
<td>0.89</td>
</tr>
<tr>
<td>( \sigma(L^u) ) (U.S. Aggregate Manufacturing)</td>
<td>1.15</td>
<td>1.15</td>
<td>1.15</td>
<td>1.15</td>
<td>1.15</td>
</tr>
<tr>
<td>( \sigma(L^i)/\sigma(L^u) )</td>
<td>2.39</td>
<td>2.42</td>
<td>1.94</td>
<td>2.08</td>
<td>2.21</td>
</tr>
<tr>
<td>( \sigma(L^m)/\sigma(L) )</td>
<td>0.77</td>
<td>0.77</td>
<td>0.77</td>
<td>0.77</td>
<td>0.77</td>
</tr>
</tbody>
</table>

| **Correlations, Employment** |         |                      |                        |                     |         |
| \( \text{corr}(L^i,L^i) \) | 0.49    | 0.43                 | 0.66                   | 0.45                | 0.51    |
| \( \text{corr}(L^u,L^u) \) | 0.78    | 0.78                 | 0.78                   | 0.78                | 0.78    |
| \( \text{corr}(L^m,L^m) \) | 0.56    | 0.71                 | 0.62                   | 0.68                | 0.64    |
| \( \text{corr}(L^u,L) \) | 0.66    | 0.86                 | 0.88                   | 0.63                | 0.76    |

**Notes:**
The top portion of the table shows standard deviations (in percent) for the production-worker employment in specific Mexico and U.S. outsourcing industries, and in Mexico and U.S. aggregate manufacturing, and the ratios of these standard deviations. Each series is in log values, seasonally adjusted, and HP filtered. Data are monthly from 1996 through 2005. The bottom portion of the table shows correlations between the various employment series.
Table 2. Size of Outsourcing Industries in Mexico and the U.S.

<table>
<thead>
<tr>
<th>NAICS</th>
<th>Industry</th>
<th>Thousands of employees (mean 2000-2005)</th>
<th>Mexico</th>
<th>U.S.</th>
<th>Texas</th>
<th>California</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>All maquiladoras (Mexico)</td>
<td></td>
<td>1,151.00</td>
<td>--</td>
<td>--</td>
<td>--</td>
</tr>
<tr>
<td></td>
<td>All manufacturing (United States)</td>
<td></td>
<td>--</td>
<td>15,336.70</td>
<td>955.5</td>
<td>1,649.00</td>
</tr>
<tr>
<td>315</td>
<td>Apparel</td>
<td></td>
<td>230.8</td>
<td>356.9</td>
<td>--</td>
<td>97.4</td>
</tr>
<tr>
<td>334</td>
<td>Computer &amp; Electronics</td>
<td></td>
<td>265.6</td>
<td>1,512.30</td>
<td>132.9</td>
<td>366.6</td>
</tr>
<tr>
<td>335</td>
<td>Electrical machinery</td>
<td></td>
<td>100.2</td>
<td>497.5</td>
<td>20.0</td>
<td>38.5</td>
</tr>
<tr>
<td>336</td>
<td>Transport equipment</td>
<td></td>
<td>240.7</td>
<td>1,855.80</td>
<td>85.2</td>
<td>137.5</td>
</tr>
</tbody>
</table>

Table 3. Relative Volatility in Mexico and U.S. Outsourcing Industries: Total Employment at the U.S. State Level

<table>
<thead>
<tr>
<th>Industry</th>
<th>Apparel</th>
<th>Electrical Machinery</th>
<th>Computer &amp; Electronics</th>
<th>Transport Equipment</th>
<th>Average</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>National Level</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$\sigma(L^*_i)$ (Mex. Outsourcing Industry)</td>
<td>4.48</td>
<td>4.11</td>
<td>5.50</td>
<td>2.73</td>
<td>4.21</td>
</tr>
<tr>
<td>$\sigma(L_i)$ (U.S. Outsourcing Industry)</td>
<td>1.63</td>
<td>1.52</td>
<td>2.47</td>
<td>1.07</td>
<td>1.67</td>
</tr>
<tr>
<td>$\sigma(L^*)$ (Mex. Aggregate Manufacturing)</td>
<td>0.77</td>
<td>0.77</td>
<td>0.77</td>
<td>0.77</td>
<td>0.77</td>
</tr>
<tr>
<td>$\sigma(L)$ (U.S. Aggregate Manufacturing)</td>
<td>1.01</td>
<td>1.01</td>
<td>1.01</td>
<td>1.01</td>
<td>1.01</td>
</tr>
<tr>
<td>$\sigma(L^*_i)/\sigma(L_i)$</td>
<td>2.75</td>
<td>2.70</td>
<td>2.23</td>
<td>2.55</td>
<td>2.56</td>
</tr>
<tr>
<td>$\sigma(L^*)/\sigma(L)$</td>
<td>0.76</td>
<td>0.76</td>
<td>0.76</td>
<td>0.76</td>
<td>0.76</td>
</tr>
</tbody>
</table>

| **California**                    |         |                      |                        |                     |         |
| $\sigma(L^*_i)$ (Mex. Outsourcing Industry) | 4.48    | 4.11                 | 5.50                   | 2.73                | 4.21    |
| $\sigma(L_i)$ (U.S. Outsourcing Industry) | 2.25    | 2.35                 | 2.62                   | 1.31                | 2.13    |
| $\sigma(L^*)$ (Mex. Aggregate Manufacturing) | 0.77    | 0.77                 | 0.77                   | 0.77                | 0.77    |
| $\sigma(L)$ (U.S. Aggregate Manufacturing) | 1.40    | 1.40                 | 1.40                   | 1.40                | 1.40    |
| $\sigma(L^*_i)/\sigma(L_i)$       | 1.99    | 1.75                 | 2.10                   | 2.08                | 1.98    |
| $\sigma(L^*)/\sigma(L)$          | 0.55    | 0.55                 | 0.55                   | 0.55                | 0.55    |

| **Texas**                         |         |                      |                        |                     |         |
| $\sigma(L^*_i)$ (Mex. Outsourcing Industry) | 4.48    | 4.11                 | 5.50                   | 2.73                | 3.09    |
| $\sigma(L_i)$ (U.S. Outsourcing Industry) | n.a.   | 2.48                 | 3.12                   | 1.66                | 2.42    |
| $\sigma(L^*)$ (Mex. Aggregate Manufacturing) | 0.77    | 0.77                 | 0.77                   | 0.77                | 0.77    |
| $\sigma(L)$ (U.S. Aggregate Manufacturing) | 1.16    | 1.16                 | 1.16                   | 1.16                | 1.16    |
| $\sigma(L^*_i)/\sigma(L_i)$       | n.a.    | 1.66                 | 1.76                   | 1.64                | 1.69    |
| $\sigma(L^*)/\sigma(L)$          | n.a.    | 0.66                 | 0.66                   | 0.66                | 0.66    |

**Notes:**

The table follows the same format as the top portion of Table 1, but uses total employment rather than production-worker employment.

n.a. indicates this industry is not available for that state.
Table 4: Adjustment in the Maquiladora Industry: Extensive Margins

<table>
<thead>
<tr>
<th></th>
<th>Number of plants</th>
<th>Employment per plant</th>
<th>No. of HS Products</th>
<th>Sales per HS Product</th>
<th>Extensive Margin of Exports</th>
<th>Intensive Margin of Exports</th>
</tr>
</thead>
<tbody>
<tr>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
<td>(6)</td>
<td></td>
</tr>
<tr>
<td>Industry share of</td>
<td>0.376</td>
<td>0.624</td>
<td>(0.163)</td>
<td>(0.163)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>aggregate employment</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Aggregate employment</td>
<td>0.494</td>
<td>0.506</td>
<td>(0.052)</td>
<td>(0.052)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Industry share of</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>exports at border</td>
<td>0.071</td>
<td>0.929</td>
<td>0.131</td>
<td>0.869</td>
<td></td>
<td></td>
</tr>
<tr>
<td>crossing (0.021)</td>
<td>(0.021)</td>
<td>(0.025)</td>
<td>(0.025)</td>
<td>(0.025)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Total exports</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>at border crossing</td>
<td>0.103</td>
<td>0.897</td>
<td>0.085</td>
<td>0.915</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(0.027)</td>
<td>(0.027)</td>
<td>(0.025)</td>
<td>(0.025)</td>
<td>(0.025)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>R²</td>
<td>0.303</td>
<td>0.513</td>
<td>0.02</td>
<td>0.77</td>
<td>0.04</td>
<td>0.70</td>
</tr>
<tr>
<td>N</td>
<td>480</td>
<td>480</td>
<td>1584</td>
<td>1584</td>
<td>1584</td>
<td>1584</td>
</tr>
</tbody>
</table>

Notes:

Columns (1) and (2) show regressions of either the number of plants or employment per plant on total Mexican manufacturing employment and the industry share of manufacturing employment. The sample is the four outsourcing industries in Mexico, with data at a monthly frequency from 1996:1 to 2005:12. Columns (3) and (4) show regressions of either the number of HS products imported by the U.S. per month, or the average import sales per HS product, on the industry share of imports at that border crossing and total imports from Mexico at that border crossing. Columns (5) and (6) show regressions of the extensive margin of Mexican exports (which is a weighted count of the number of HS products) and the intensive margin of Mexican exports (which is the industry sales divided by the extensive margin), on the same independent variables. The sample is the four outsourcing industries in Mexico, exporting to three land border crossings, with data at a monthly frequency from 1996:1 to 2006:12. All variables are in logs, expressed in real terms, deseasonalized, and HP filtered. All regressions include controls for industry fixed effects, which are not shown. Standard errors (clustered by industry) are in parentheses.
Table 5: U.S. Harmonized System Imports from Mexico, 1996–2006

<table>
<thead>
<tr>
<th></th>
<th>Apparel</th>
<th>Electrical Machinery</th>
<th>Computer &amp; Electronics</th>
<th>Transport Equipment</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Laredo, TX</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mean Number of HS Products</td>
<td>384.0</td>
<td>219.2</td>
<td>258.6</td>
<td>140.5</td>
</tr>
<tr>
<td>Mean Number of Months a HS Product is Imported Per Year(^a)</td>
<td>6.9</td>
<td>8.9</td>
<td>7.5</td>
<td>9.0</td>
</tr>
<tr>
<td>Std. Dev. Log Number of HS Products (^b)</td>
<td>2.64</td>
<td>2.14</td>
<td>3.10</td>
<td>2.94</td>
</tr>
<tr>
<td>Correlation of Number of HS Products and U.S. Manufacturing Employment (^c)</td>
<td>0.28</td>
<td>0.32</td>
<td>0.07</td>
<td>0.31</td>
</tr>
<tr>
<td><strong>El Paso, TX</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mean Number of HS Products</td>
<td>162.6</td>
<td>137.0</td>
<td>208.3</td>
<td>53.4</td>
</tr>
<tr>
<td>Mean Number of Months a HS Product is Imported Per Year(^a)</td>
<td>5.8</td>
<td>8.9</td>
<td>7.8</td>
<td>7.8</td>
</tr>
<tr>
<td>Std. Dev. Log Number of HS Products (^b)</td>
<td>4.99</td>
<td>2.43</td>
<td>2.60</td>
<td>5.66</td>
</tr>
<tr>
<td>Correlation of Number of HS Products and U.S. Manufacturing Employment (^c)</td>
<td>0.05</td>
<td>0.04</td>
<td>0.12</td>
<td>0.30</td>
</tr>
<tr>
<td><strong>San Diego, CA</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mean Number of HS Products</td>
<td>299.8</td>
<td>129.6</td>
<td>237.0</td>
<td>53.2</td>
</tr>
<tr>
<td>Mean Number of Months a HS Product is Imported Per Year(^a)</td>
<td>6.3</td>
<td>8.3</td>
<td>7.5</td>
<td>7.0</td>
</tr>
<tr>
<td>Std. Dev. Log Number of HS Products (^b)</td>
<td>3.38</td>
<td>3.86</td>
<td>3.39</td>
<td>6.18</td>
</tr>
<tr>
<td>Correlation of Number of HS Products and U.S. Manufacturing Employment (^c)</td>
<td>-0.01</td>
<td>0.28</td>
<td>0.25</td>
<td>0.12</td>
</tr>
</tbody>
</table>

**Notes:**

- Averaged over HS products and over the years 1996–2006.
- The log number of HS products has been deseasonalized and HP filtered, and the standard deviation is multiplied by 100.
- The number of HS products and U.S. manufacturing employment are in logs, and are deseasonalized and HP filtered.

**Source:**

Table 6. Calibration of model Parameters

### Preferences

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\theta$</td>
<td>home bias in U.S.</td>
<td>0.88</td>
</tr>
<tr>
<td>$\theta^*$</td>
<td>home bias in Mexico</td>
<td>0.71</td>
</tr>
<tr>
<td>$\alpha$</td>
<td>outsourcing expenditure share</td>
<td>0.24</td>
</tr>
<tr>
<td>$\sigma$</td>
<td>elasticity between varieties</td>
<td>9.5</td>
</tr>
<tr>
<td>$\eta$</td>
<td>elasticity, multinational and domestic goods</td>
<td>1</td>
</tr>
<tr>
<td>$\chi$</td>
<td>elasticity, home and foreign goods</td>
<td>2</td>
</tr>
<tr>
<td>$\mu$</td>
<td>labor supply elasticity</td>
<td>1</td>
</tr>
<tr>
<td>$\phi$</td>
<td>risk aversion</td>
<td>2</td>
</tr>
<tr>
<td>$n$</td>
<td>relative size of U.S.</td>
<td>0.74</td>
</tr>
<tr>
<td>$\sigma_r$</td>
<td>relative wealth of U.S.</td>
<td>24.5</td>
</tr>
<tr>
<td>$\bar{G}$</td>
<td>US mean government demand</td>
<td>0.186</td>
</tr>
<tr>
<td>$\bar{G}^*$</td>
<td>Mexican mean government demand</td>
<td>0.0432</td>
</tr>
</tbody>
</table>

### Technology

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\bar{a}_H$</td>
<td>US steady state unit cost</td>
<td>1</td>
</tr>
<tr>
<td>$\bar{a}_F$</td>
<td>Mexican steady state unit cost</td>
<td>4.22</td>
</tr>
<tr>
<td>$\bar{\varepsilon}$</td>
<td>outsourcing margin</td>
<td>0.06</td>
</tr>
<tr>
<td>$a$</td>
<td>US outsourcing slope parameter</td>
<td>-0.609</td>
</tr>
<tr>
<td>$\bar{b}$</td>
<td>US outsourcing level parameter</td>
<td>0.0365</td>
</tr>
<tr>
<td>$a_d$</td>
<td>relative cost slope parameter</td>
<td>-2.14</td>
</tr>
<tr>
<td>$\bar{b}_d$</td>
<td>relative cost level parameter</td>
<td>-1.95</td>
</tr>
</tbody>
</table>

### Shock processes

$$\begin{align*}
\sigma_a &= \begin{bmatrix}
6.42 \times 10^{-5} & -4.67 \times 10^{-6} \\
-4.67 \times 10^{-6} & 1.87 \times 10^{-4}
\end{bmatrix} \\
\rho_a &= \begin{bmatrix}
0.931 & -4.02 \times 10^{-2} \\
7.38 \times 10^{-3} & 0.961
\end{bmatrix} \\
\sigma_G &= \begin{bmatrix}
7.68 \times 10^{-5} & 4.02 \times 10^{-3} \\
4.02 \times 10^{-3} & 2.36 \times 10^{-2}
\end{bmatrix} \\
\rho_G &= \begin{bmatrix}
-0.0549 & 0.379 \\
-3.68 & 0.424
\end{bmatrix}
\end{align*}$$
Table 7. Model Simulation for Production Worker Employment in the Outsourcing Sector

<table>
<thead>
<tr>
<th></th>
<th>(1)</th>
<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
<th>(6)</th>
<th>(7)</th>
<th>(8)</th>
<th>(9)</th>
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<td>$cov(z',D_W/W^*)^a$</td>
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Notes:

a All variables are in logs except for $z'$ in this expression.
b $A'(z) = -0.1$.
c Holding the slope $A'(z)$ constant at its benchmark value of -0.2677.
d $A(z)$ distribution shifted by productivity shocks.
Figure 1: Maquiladora Activity in Mexico
(log values)
### Figure 2: Employment for Production Workers in Mexico and U.S. Outsourcing Industries
(log values, seasonally adjusted and HP filtered)

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Figure 3: The Number of HS Products Over Years
(Ave. over 3 ports; Log values, seasonally adjusted, HP filtered)