

Offshoring Jobs? Multinationals and US Manufacturing Employment *

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December 2007

*For assistance with data the authors would like to thank Raymond Mataloni, Fritz Foley and Stanley Watt. For helpful comments we wish to thank David Card, the fellows at the Radcliffe Institute for Advanced Study, Larry Katz, and Francis Kramarz, seminar participants at the BEA, Columbia University, the University of Michigan, Stanford, UC Berkeley, Yale University, the IMF, the Paris School of Economics, the University of Maryland and Pierluigi Balduzzi. For financial assistance the authors gratefully acknowledge the National Science Foundation. McMillan acknowledges the Radcliffe Institute for Advanced Study for both financial support and time to devote to this project. The statistical analysis of firm-level data on US multinational companies was conducted at the International Investment Division, Bureau of Economic Analysis of the US Department of Commerce under arrangements that maintain legal confidentiality requirements. The views expressed in this paper are those of the authors and do not reflect official positions of the US Department of Commerce. Research assistance from Joan Hamory, Clair Null, and Andrew Waxman is gratefully acknowledged.

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Abstract:

Critics of globalization claim that US manufacturing firms are being driven to shift employment abroad by the prospects of cheaper labor. Others argue that the availability of low-wage labor has allowed US-based firms to survive and even prosper. Yet evidence for either hypothesis, beyond anecdotes, is slim. Using firm-level data collected by the US Bureau of Economic Analysis (BEA), we estimate the impact on US manufacturing employment of changes in foreign affiliate wages, controlling for changing demand conditions and technological change. We find that the evidence supports both perspectives on globalization. For firms most likely to perform the same tasks in foreign affiliates and at home (“horizontal” foreign investment), foreign and domestic employees appear to be substitutes. For these firms, lower wages in affiliate locations are associated with lower employment in the US. However, for firms which do significantly different tasks at home and abroad (“vertical” foreign investment), foreign and domestic employment are complements. These offsetting effects may be combined to show that offshoring is associated with a quantitatively small decline in manufacturing employment. In fact, for vertical multinationals, offshoring is associated with significant employment gains. Other factors, such as declining prices for consumer goods, import competition, and falling prices for investment goods (which substitute for labor) play a more important role in explaining the contraction in US manufacturing employment.

I. Introduction

During the last three decades, domestic manufacturing employment of US-based multinationals has fallen steadily.¹ Between 1982 and 1999, foreign employment of US multinationals increased, climbing from 26 percent to nearly 40 percent of their labor force. These parallel developments have led critics of globalization to conclude that US firms are cutting employment at home and shifting employment abroad. Concerns about offshoring have intensified as newly released data indicates a further decline in manufacturing employment both by US-based multinationals and for the US economy as a whole.

The public outcry motivated Congress to take action. On October 22, 2004 the US Congress passed the American Jobs Creation Act of 2004. The Act contains a provision to encourage profit repatriation back to the US by domestic multinationals--explicitly for the purpose of job creation at home. Yet the evidence linking offshore activities to falling domestic labor demand is, in fact, contradictory. Several studies suggest that there are no employment losses from offshoring activities. The first important research to illustrate this point is Brainard and Riker (1997), who showed that employment in low and high income affiliates of US multinationals is complementary for manufacturing activities. Borga (2005) and Desai, Foley, and Hines (2005) also find that expansion of US multinationals abroad *stimulates* job growth at home. Reviewing these studies, Mankiw and Swagel (2006) conclude that “foreign activity does not crowd out domestic activity; the reverse is true.”

A second set of studies (Brainard and Riker (2001), Hanson, Mataloni and Slaughter (2003), Muendler and Becker (2006)) reaches the opposite conclusion: jobs abroad *replace* jobs at home, but the effect is small. These different answers are problematic for policymakers who are left uncertain about how to respond to the growing ranks of US firms setting up shop abroad. How can we reconcile these two different sets of findings? We believe that the answer lies in understanding the different motivations for foreign investment.

¹ See Table 1, which shows a four million worker decline between 1982 and 1999.

Theoretical models of trade and foreign investment imply that different types of foreign investments will be associated with opposite effects on domestic labor demand. Markusen and Maskus (2001) separate multinationals into vertically- and horizontally-integrated types. They show how different incentives for foreign investment lead to different organizational structures, which in turn should produce different degrees of substitution between employment at home and abroad. Horizontal multinationals, which are defined as firms which produce the same products in different locations, are primarily motivated by trade costs to locate abroad.² For these types of firms, investment abroad substitutes for parent exports. One implication is that foreign affiliate employment should substitute for home employment. For vertically-integrated firms, however, trade and foreign investment are complements. Vertically-integrated enterprises are motivated by factor endowment differences (and consequently factor price differences in a world where there is not factor price equalization) to locate different components of production in different locations. As pointed out by Brainard and Riker (1997), one implication of this kind of multinational activity is that parent and affiliate employment should be complementary.

More recent work by Grossman and Rossi-Hansberg (2006) draws on insights from Autor, Levy and Murnane (2002) to develop a framework in which falling costs of offshoring can lead to wage gains for workers at home. Grossman and Rossi-Hansberg (2006) use Autor, Levy and Murnane's differentiation between routine and non-routine tasks to build a theoretical model of trade in tasks. Advances in technology (such as improvements in communication) make offshoring of routine tasks less costly, leading firms to increase production abroad. What is surprising is that offshoring of routine tasks for vertically-integrated multinationals (there is no horizontal motive for foreign investment here) leads to ambiguous predictions for domestic wages. The intuition behind this result is that falling costs of offshoring act like a positive productivity shock. Although the primary motivation for offshoring is to reduce labor costs, low-skill workers at home may still gain if terms of trade effects and labor supply effects are not too large.

In this paper, we develop an empirical framework which is flexible enough to allow substitution or complementarity between home and affiliate employment for firms that have different motivations to

² For the purpose of simplicity, we will occasionally refer to horizontally-integrated firms as horizontal firms, and

engage in foreign activities. With this framework, we are able to identify the separate effects of horizontal versus vertical foreign investment on home employment, and also allow for different degrees of substitution (or complementarity) in high- and low-income affiliate locations. At the same time, we control for other confounding changes, such as other factor price changes, demand shocks, and technological change. To address the possibility that methodological differences might be driving the conflicting results described above, we adopt a variety of different approaches to estimating labor demand and a range of econometric techniques.

We find that the insights derived from trade theory go a long way towards explaining the apparently contradictory evidence on the relationship between offshoring and domestic manufacturing employment. For US parents involved in primarily horizontal activities, affiliate activity abroad substitutes for domestic employment. For vertically-integrated parents, however, the results suggest that home and foreign employment are complementary. Foreign wage reductions are associated with an increase in domestic employment. The results differ across high- and low-income affiliate locations, in part because factor-price differences relative to the US are much more important in low-income regions. In low income affiliate locations, a 10 percentage point reduction in wages is associated with 2.7 percent reduction in US parent employment for horizontal parents and a 3.1 percentage point increase in parent employment for vertical firms.

We also show that offshoring is not the primary driver of declining domestic employment of US manufacturing multinationals between 1977 and 1999. In fact, the evidence suggests that operating in low-income affiliate locations preserves jobs (for vertically integrated parents), instead of destroying them. We show that declining domestic employment of US multinationals is primarily due to falling prices of investment goods (such as computers, which substitute for labor), falling prices of consumption goods, and increasing import competition. Our research highlights both the importance of heterogeneous firm responses to opportunities for direct investment abroad and the need to account for other avenues through which international competition affects US labor demand.

Our results are consistent with the literature that focuses on the impact of international trade on US jobs. Revenga (1992) finds a negative impact of changes in import prices on US employment growth.

vertically-integrated firms as vertical.

Katz and Murphy (1992) also find that increased import competition negatively affected relative labor demand in the US, particularly in the 1980s with the growing of the US trade deficit. Borjas, Freeman and Katz (1997) find that increased trade with developing countries depresses wages at the bottom of the income distribution. Bernard, Jensen and Schott (2006) examine the impact of US imports on both the survival and employment of US manufacturing firms. They find that imports only harm US manufacturing employment when those imports are from low wage countries.

The remainder of this paper is organized as follows. In Section II, we describe the Bureau of Economic Analysis data on outward direct investment and our choice of sample. Section III describes the empirical framework and discusses econometric issues. Section IV presents the results and Section V concludes.

II. The BEA Data

We analyze firm-level surveys on US direct investment abroad, collected each year by the Bureau of Economic Analysis (BEA) of the US Department of Commerce. The BEA collects confidential data on the activities of US-based multinationals, defined as the combination of a single US entity that has made the direct investment, called the parent, and at least one foreign business enterprise, called the foreign affiliate. We use the data collected on majority-owned, non-bank foreign affiliates and non-bank US parents for the benchmark years from 1982 and 1999. These benchmark years (1982, 1989, 1994 and 1999) include more comprehensive information than the annual surveys.³

Creating a panel using the benchmark years of the BEA survey data is a nontrivial task for several reasons. First, not all firms are required to report to the BEA and reporting requirements vary across years. Second, we must consider the implications of the changes to the Standard Industrial Classification (SIC) codes in 1972 and 1987 and the switch from SIC codes to the North American Industrial Classification System (NAICS) codes in 1997. The fact that parents are allowed to consolidate information for several affiliates in one country on a single form calls for special care in the aggregation and interpretation of affiliate level data.

All foreign affiliates with sales, assets or net income in excess of a certain amount in absolute value must report their data to the BEA. This amount was \$3 million dollars in 1982, 1989 and 1994 and rose to \$7 million dollars in 1999. In addition, a new reporting requirement was imposed on parents in 1999. Parents whose sales, assets or net income exceeded \$100 million (in absolute value) were required to provide more extensive information than parents whose sales, assets or net income fell below that level.⁴ To determine whether the changes in reporting requirements made small firms overrepresented in our sample in the early years, we imposed a double filter on the data using the uniform cutoff for affiliates (based on the strictest reporting requirement of \$100 million in 1999) of \$5.59 million in 1982 US dollars and \$79.87 1982 US dollars for parents. As it turns out, the reporting requirements were large enough that imposing the filter on the data makes little difference on our initial results. Therefore, we use all of the available data.

Finally, we face selection issues with our sample of “manufacturing” firms.⁵ We keep those parents whose primary industry of sales is manufacturing since our goal is to determine whether manufacturing jobs at home are being replaced by manufacturing jobs abroad. However, some parents were reclassified from manufacturing to wholesale trade and services. To account for this, we keep all parents that were *ever* classified in manufacturing and their manufacturing affiliates.⁶

³ While the BEA collects annual data on US direct investment abroad, these data do not include all the variables we need and can find in the benchmark years.

⁴ Parents who do not meet this cutoff but who have affiliates that meet the \$7 million cutoff are still required to provide extensive information for affiliates.

⁵ To document what has happened within industries in manufacturing over time, we created a concordance that allows us to assign SIC codes to NAICS codes. This was necessary because in 1999 the BEA collected data on NAICS codes and not SIC codes. We chose to convert SIC codes to NAICS codes since all future information will be collected on the basis of NAICS codes. For example, data for the benchmark year 2004 will be available shortly and firms report based on NAICS codes. The 1977 and 1982 benchmark years are based on the 1972 SIC codes. The 1989 and 1994 benchmark years are based on the 1987 SIC codes. The 1999 benchmark data are based on the 1997 NAICS codes. In addition to the fact that the industry codes are not directly comparable across all benchmark years, the BEA industry codes have been slightly modified to reflect the fact that these are enterprise data and are called, respectively, SIC-ISI and NAICS-ISI. Working with these codes, we created a program (available upon request) that assigns the SIC-ISI codes for the years 1977-1994 to NAICS-ISI codes. Both parents and affiliates are classified into their primary industry of sales using the following algorithm, which tracks the algorithm used by the BEA: the top five industries by parent or affiliate sales are used to assign to each parent or affiliate one of the 22 aggregates. Sales are collapsed into the top five industries of sales and then the maximum sale by industry is identified. A parent or affiliate is classified as being in manufacturing if its maximum sales across the top five industries of sales is in manufacturing.

⁶ There are a number of parents who have been reclassified from manufacturing to wholesale trade and services. For example, several firms were in manufacturing but are now classified in wholesale trade because almost all of their manufacturing is done overseas and not in the United States. To account for this, we chose our sample in two different ways. First, we included parents who either were classified in manufacturing or had previously been classified in manufacturing and their manufacturing affiliates. Next, we included only parents who were currently in manufacturing in any given year and their manufacturing affiliates. Since the results are not sensitive to this

Table 1 reports the number of employees of US manufacturing parents both in the US and in foreign affiliate locations. US employment of manufacturing parents declined from nearly 12 million in 1982 to slightly below 8 million in 1999. The second row of Table 1 shows the employment coverage of our sample after we perform the cleaning procedures described above. The sample size remains almost the same, particularly in the later years. The following two rows report the breakdown of US employment for horizontal versus vertical multinationals (discussed in more detail below and defined in Table 2). While US employment of both types of multinationals declined, foreign affiliate employment increased. Almost all the increase in foreign affiliate employment occurred in low-income affiliate locations, and affiliate employment gains in vertically-integrated firms were twice as large as for horizontally-integrated firms.

The share of US multinational employment concentrated in affiliates increased from 26 percent in 1982 to 39 percent in 1999. Although total affiliate employment increased by more than one million employees, the foreign employment gains did not fully offset the domestic losses. This suggests that there are other important determinants of falling domestic employment for US multinationals. Alternative explanations, which shall be incorporated into our empirical framework, include changing prices of capital, labor-saving technical change, and increased import competition.

Manufacturing multinationals reporting to the BEA accounted for the majority of economic activity in US manufacturing during the sample period. Appendix Table A.1 (based on Mataloni and Fahim-Nader (1996) and Mataloni and Yorgason (2006)) reports the coverage of the BEA data for benchmark years 1982 through 1999. In 1982, gross product by these enterprises accounted for over 80 percent of total manufacturing and 77 percent of manufactured exports in the United States. By 1999, the BEA's coverage had declined slightly: these enterprises accounted for only 63 percent of US exports and about half of manufacturing employment. These firms also accounted for more than 80 percent of total private US research and development expenditures throughout the sample period (Mataloni and Fahim-Nader (1996)). Appendix Table A.1 also shows that the proportion of services firms accounted for by the BEA sample is extremely small. During the sample period, the BEA sample accounted for only between

distinction, we use the larger of the two samples, keeping all parents that were *ever* classified in manufacturing and their manufacturing affiliates.

6 and 8 percent of total gross product in services. Consequently, we restrict our analysis to manufacturing, which we believe provides a more representative sample.

How reliable are these data? These are the only data officially collected by a US government agency on affiliate activity abroad. We have initiated a number of data checks to analyze the reliability of the coverage.^{7,8} We were able to cross-check the employment numbers for US affiliate activity reported by the BEA with data on inward foreign investment reported by the official statistical agencies in Germany and Sweden. These checks are reported in Appendix Table A.2. We report total employment in both countries as indicated by the BEA database and show that it is quite close to the same numbers collected by the national statistical agencies. Although there are some discrepancies between BEA and German and Swedish data, this may be, at least partially, accounted for by variation in reporting based upon fiscal year versus calendar year. The BEA classifies a firm in 1999 if its fiscal year ends in 1999—this could be for any month in 1999. Although most firms have their fiscal year ending in December, enough have earlier end dates that some of the 1999 BEA employment figures correspond to a mix of the 1998 and 1999 employment figures reported by the statistical bureaus for Sweden and Germany.

III. Empirical Framework

Previous work has used very different econometric models to specify the impact of foreign affiliate activity on labor demand at home, making it difficult to identify whether the conflicting results stem from different approaches or different datasets and time periods. Brainard and Riker (1997) estimate labor demand as a function of wages in different locations, Desai et al. (2006) estimate a reduced form equation with log labor at home as a function of log labor abroad, and Brainard and Riker (2001), Hanson, Mataloni and Slaughter (2003) and Muendler and Becker (2006) use a short-run translog cost

⁷ We are particularly grateful to Marc Muendler and Karolina Eckholm for helping us do this cross-checking. They provided the data on the activities of US multinational affiliates in Germany and Sweden.

⁸ We also contacted Statistics Canada to check whether they record information on affiliates of US multinationals in Canada, which would allow us to cross-check US data on foreign affiliates there with Canadian data on inward

function approach to derive factor shares as a function of wages in different locations. Katz and Murphy (1992) and Card (2001), focusing on the effects of immigration and trade, both use a CES functional form to derive an equilibrium relationship between the ratio of employment at home to employment abroad and the ratio of wages at home to wages abroad.

We chose as our primary specification to derive labor demand from a generalized cost function. Our preferred approach is attractive for several reasons. It puts minimal restrictions on the nature of the production function, unlike the CES specification which imposes a constant elasticity of substitution across different factor inputs. In the translog approach, we worry that identifying elasticities of substitution or complementarity and calculating standard errors is a less transparent process (depending, among other things, on the choice of factor shares) than estimating a labor demand equation. Previous approaches in the offshoring literature have imposed a short-run cost function and kept capital inputs fixed, which would make it difficult for us to compare our results to earlier coefficient estimates. However, for completeness we also derive estimating equations using a generalized translog and CES function approach. We shall see that the implied elasticities of complementarity (or substitution) are remarkably robust across these different specifications.

Modifying Hamermesh (1993), let us consider a firm using N domestic factors and N^* foreign factors of production $X_1 \dots X_N, X_{1^*} \dots X_{N^*}$. We begin by assuming there are only two locations (domestic and foreign) but will generalize to j locations in the empirical specification which follows. Let the production function for a US multinational firm i producing total aggregate worldwide output Y_i and using N domestic and N^* foreign inputs X_i and X_{i^*} be

$$(1) \quad Y_i = f(X_{1i}, \dots, X_{Ni}, X_{1i^*}, \dots, X_{N_i^*}), f_i > 0, f_{ii} < 0$$

Output Y can include production at home and abroad and production could be exported or sold on domestic markets. Then the associated cost function, based on the demands for X_1 through X_N and X_{1^*} through X_{N^*} is given by

$$(2) \quad C_i = g(w_{1i}, \dots, w_{Ni}, w_{1i^*}, \dots, w_{N_i^*}, Y_i), g_i > 0$$

foreign investment. Statistics Canada informed us that they do not gather data on affiliates because it is too difficult to define a foreign affiliate and referred us to the BEA.

where the w_i 's and w_{i^*} 's are the N and N* input prices at home and in the foreign affiliate location. One can use Shepard's lemma to derive the factor demand for the n th input for US multinational firm i :

$$(3) \quad X_{ni} = X_{ni}^d(w_{1i}, \dots, w_{Ni}, w_{1i^*}, \dots, w_{Ni^*}, Y_i), n = 1, \dots, N, n^* = 1, \dots, N^*$$

Our first approach will be to estimate a log-linear version of equation (3), focusing on US labor demand and extending (3) to allow for three locations. With three locations, there are three types of labor inputs: home labor, foreign labor in low-income affiliates, and foreign labor in high-income affiliates. This framework is flexible enough to allow for a range of production technologies, including Brainard and Riker's (1997) assumption that production is vertically decomposed across high-wage and low-wage regions. We will also allow for two other types of inputs, making the total number of inputs N in each location equal to three: labor, physical capital, and research and development inputs. As with wages, we allow physical capital and research and development inputs to be separately identified depending on location.

One estimation issue which arises is that global output Y for firm i is jointly determined with domestic US employment. If we were to estimate (3) directly, we would have a significant simultaneity problem. We solve this by assuming that aggregate worldwide output Y for firm i is a function of domestic and foreign prices:

$$Y_i = Y(P, P^*)$$

Substituting this into (3) yields:

$$(4) \quad X_{ni} = X_{ni}^d(w_{1i}, \dots, w_{Ni}, w_{1i^*}, \dots, w_{Ni^*}, P, P^*), n = 1, \dots, N, n^* = 1, \dots, N^*.$$

Our first set of estimating equations is based on log-linearization of (4), generalizing to j locations, and takes the following form:

$$(5) \ln L_{it} = \beta_0 + \sum_j \alpha_j \ln P_{jt} + \sum_j \eta_j w_{ijt} + \sum_j \omega_j r_{ijt} + \sum_j \chi_j t_{ijt} + d_t + f_i + \varepsilon_{ijt}.$$

The dependent variable $\ln L$ is the natural logarithm of net annual employment by the US parent in the United States, the P 's are final goods prices, w is the wage in location j and time t , r is the price of capital in location j and time t , and t is the price of research and development goods. We allow for time effects d and a firm-specific (common to the parent and its affiliate) fixed effect f_i , which takes into account both firm-specific productivity differences and other non-varying firm characteristics, while j indexes location and t indexes time.

Defining Horizontal and Vertical Multinationals

Markusen and Maskus (2001), in their comprehensive survey of general equilibrium approaches to the multinational firm, define horizontal multinationals as “firms that produce the same product in multiple plants, serving local markets by local production.” This definition of horizontal integration implies that intra-firm trade will be low, since foreign investment substitutes for US exports. Vertical firms are defined as “firms that fragment the production process into stages based on factor intensities and locate activities according to international differences in factor prices.” An important finding of Markusen and Maskus is that foreign investment replaces trade in the case of horizontal multinationals but is positively correlated with trade in the case of vertical foreign investment. Although we cannot directly test the motivation for foreign investment with our data, we can use the different implications for intra-firm trade as a way to distinguish vertical from horizontal foreign investment.

We construct intra-firm trade as the sum of exports to foreign affiliates for further processing plus imports from foreign affiliates as a share of sales. We then define firms as vertical if they have a high share of intra-firm trade and firms as horizontal if they have a low share. Grossman and Rossi-Hansberg (2006) also use intra-firm trade to quantify the increase in vertical activities of multinationals, pointing out that intra-firm trade “mostly reflects the international division of labor within multinational enterprises.” We measure high intra-firm trade in two ways. First we run a regression of intra-firm trade on industry dummies, and define firms as vertical if the industry dummy is statistically significant. As

reported in Table 2, this is equivalent to selecting those industries with a high average share of intra-firm trade during the sample period as vertical multinationals.

The results in Table 2 suggest that the following industries may be characterized as vertical: chemicals, plastics, primary and fabricated metals, machinery, computers and electronics, electrical equipment, and transportation equipment. The remaining industries are classified as horizontal: textiles and apparel, food, beverages and tobacco, leather products, wood and paper products, petroleum products, non-metallic minerals and furniture. While it may be surprising that apparel is classified as horizontal and not vertical, potential vertical activity in that sector is limited by high trade costs. Much horizontal FDI is motivated by trade barriers (such as tariffs or quotas). Textiles and apparel and beverages and tobacco are typically the most protected sectors in both industrial and developing countries (for developing countries, see Hanson and Harrison (2001)). To make this point more clearly, we have also listed in the last three columns a summary measure of tariffs for China, an average for all developing countries, and tariffs in the United States. The average tariffs by sector for developing countries and China is for 1999 and is taken from the World Bank's trade database. The measure of tariffs for the US includes tariffs and transport costs for each sector for the US in 1979, based on Bernard, Jensen, and Schott (2006). For the United States, we selected an earlier year to indicate more clearly the differences in the pattern of protection across sectors; in more recent years, tariffs in the United States have been reduced to very low levels. Trade frictions are highest for textiles and apparel, beverages and tobacco, leather, and non-metallic minerals (cement). This implies that firms in highly protected sectors (textiles and apparel) or in sectors with high costs of transportation (cement) must frequently engage in horizontal investments in order to access domestic foreign markets.⁹

⁹ It is also instructive to compare our classification of horizontal versus vertical with the classification used by Autor, Levy, and Murnane (2003) to categorize workers into routine and non-routine tasks. They argue that computers are likely to substitute for labor in the case of routine tasks, which could be either cognitive or manual, and most likely to be complementary with labor in the case of cognitive non-routine tasks. As pointed out by Grossman and Rossi-Hansberg (2006), "routine" tasks are more likely to be offshored via vertical foreign investment (there is no motivation for horizontal foreign investment in their framework). We posit that the type of routine tasks that are important in manufacturing are manual rather than cognitive. We pick a measure of non-routine manual tasks which measures eye-hand-foot coordination (EFT). It is possible to use the Autor, et al. data to classify industries according to their EFT level, by merging CPS worker and occupation data with their industry affiliation. In Table 2, firms defined as vertically-integrated on the basis of high intra-firm trade measure poorly in the EFT scale. So our measure of vertical versus horizontal could also be interpreted as a measure of distance from non-routine manual tasks.

Data and Estimation Issues

To estimate equation (5) we need data on US employment, capital prices, wages, final goods prices, and research and development prices. We need factor prices and output prices for each of the j locations in which the multinational firm has operations. We measure US employment as the (log) number of individuals employed by the parent in the US, since hours or even employment broken down by skill levels are not collected for US parents. Domestic prices of investment are defined at the disaggregated industry level and are taken from the NBER's manufacturing database. Domestic US wages are computed at the industry level using both the BEA and UNIDO datasets (see discussion below).

While in principle there could be as many factor and final goods prices as there are countries in the BEA database, in practice the number of j locations is limited by data availability and the need for parsimony in estimation. We restrict our j locations to 3: domestic (US) activity, high income locations, and low-income locations. One problem is that many firms, especially small enterprises, do not have any operations in low-income countries. To permit us to include these firms in the estimation, we set wages for these firms equal to zero and add a dummy variable indicating whether or not the firm has a missing observation for low-income affiliates.

For foreign prices P^* we use the prices of consumption goods, taken from the Penn World Tables. We also rely on the Penn World Table measures of foreign investment prices to capture the impact of the price of foreign capital. While in principle all foreign factor prices should be broken down into low-income foreign and high-income foreign affiliate locations, collinearity in investment and consumption prices have led us to aggregate these prices across foreign affiliate locations. Because both capital and goods are significantly more mobile than individuals, the factor price differentials across high and low-income affiliate locations are much larger for labor inputs.

In US manufacturing, international competition plays an important role in price determination, so we proxy for US industry-level prices using both industry dummies and import competition. These data were made available at the 4-digit ISIC level by Bernard, Jensen, and Schott (2006). We also include a measure of import penetration from low-wage countries, also computed by these authors. Alternative

specifications reported later in the paper use industry-level aggregates of output as alternative measures of output shocks, with separate industry-level measures of Y calculated for parents and affiliates.

We do not have adequate measures of prices for research and development goods. However, we believe that these are important inputs into production and could account for a significant impact on manufacturing employment, particularly if research and development inputs are associated with labor-saving technical change. Consequently, we proxy for prices of research and development goods with the share of research and development employment in total employment at the firm level. These are separately available for both the parent and affiliates in each location.

Since wages are calculated at the country level using BEA aggregates of the firm-level measures, we assume that wages are exogenously determined. However, we also test for the validity of this assumption by using wages collected by UNIDO. Our wage measure is defined using the following formula:

$$w_{ijt} = \sum_{c \in j} W_{ct} \frac{L_{c,t0}}{\sum_c L_{c,t0}}$$

The wage w is the wage facing the i th firm at time t in country c in affiliate location j , and the W_{ct} 's are country-time specific wages either computed using the BEA data on worker compensation aggregated to the country level or the UNIDO wages. Wages are employment-weighted averages of wages in high- and low-income affiliates, where the weights are given by the parent's initial share of employment within countries belonging to each high- and low-income category. We use initial period employment weights to avoid endogeneity problems. Affiliate country locations are defined as either high- or low-income based on the World Bank's country classifications (see Table 3).

IV. Results

We report sample means in Table 4. The US labor expenditure share in worldwide parent activity averaged 83.4 percent for horizontally-integrated and 80.7 percent for vertically-integrated firms. Affiliate labor expenditure shares accounted for between 14 and 17.2 percent of expenditures, while low-income

affiliate shares only accounted for the remaining two percent. During this period, the US parent share of worldwide labor expenses fell 4 percent, with most of the 4 percent drop in US parent shares going to low-income affiliate locations.

US parent employment shares fell considerably more than expenditure shares, falling from 74 to 61 percent of worldwide parent employment (see Table 1). The reason why the employment changes were large but expenditure share changes were small is because wage trends offset the employment developments: real wages in the sample went up in the United States but fell in high-income and low-income affiliate countries. The real wage declines abroad were large, particularly in low-income locations. One explanation which is consistent with these wage trends is a change in the composition of employment: US parents (and their high-income affiliates) are retaining relatively high-skilled workers and shifting relatively low-skilled jobs to low-income countries where labor is less expensive. This was particularly true among vertical multinationals, where domestic wages increased by 17 percent in real terms over the sample period and low-income affiliate real wages fell by 27 percent.

Research and development employment as a share of total parent employment averaged 3.1 percent for horizontal and 9.1 percent for vertically-integrated US multinationals. The significantly higher R&D shares and much larger domestic (US) wage increases for vertical multinationals is consistent with the hypothesis that these types of firms divide their activities between foreign and domestic locations, performing the most skill-intensive activities at home. R&D employment as a share of total employment rose in the US but fell in both high- and low-skill affiliate locations, suggesting that US parents chose to increasingly concentrate R&D activity in the United States.

Average import penetration in the four-digit SIC sector over the period was 12.7 percent for horizontal firms and 17.8 percent for vertical firms. Import penetration increased by 8.2 percentage points for horizontal firms and 14.6 percentage points for vertical firms, which reflects an enormous increase in the exposure of US manufacturing firms to import competition. Import competition from low-wage countries also increased for both types of enterprises, increasing by 9.9 for horizontal firms and 6.2 percentage points for vertical firms. While the price of investment goods in the US fell for both sets of enterprises, prices fell faster for vertically-integrated firms, reflecting, in part, the importance of falling computer-related costs for these firms. The real price of consumption goods, as reported by the Penn

World Tables, also fell for both sets of enterprises reflecting not just changes in demand but also advances in productivity and increased competition from lower wage countries.

Fixed Effect Results for Labor Demand

We report the results of estimating equation (5) in Table 5. The log of US employment is our dependent variable and we use a within transformation of the data to eliminate firm fixed effects. All specifications include time dummies to control for year-specific shocks. The first column of Table 5 reports coefficient estimates when we pool horizontal and vertical enterprises. The results suggest that employees in low-income affiliates are substitutes for parent home employment and employees in high-income affiliates are complements. The point estimate of 0.067 on low-income affiliate wages indicates that a 10 percent fall in foreign wages would lead to a 0.67 percent fall in US parent employment. The point estimate on high-income affiliate wages suggests the opposite: that a 10 percent increase in high-income affiliate wage increases would be associated with a 0.31 percent fall in parent employment. In columns (2) and (3), we allow the slope coefficients of our explanatory variables to vary according to whether firms are classified as vertical or horizontal according to Table 2. Formal F-tests of equality of coefficients are presented in column (4); in many cases, we reject that the coefficients are the same across the two types of enterprises. Consequently, we separately report results for vertically and horizontally integrated parents throughout the remainder of the paper.

Column (2) reports the results for vertically-integrated multinational enterprises. The coefficient on low-income affiliate wages, at -0.318, suggests that a 10 percent fall in affiliate wage levels would lead to a 3.18 percent increase in domestic (US) labor demand. For these types of firms, employment in low-income affiliates is complementary with home employment. The coefficient on high-income wages, while also negative, is smaller in magnitude and statistically insignificant. These results suggest that falling wages abroad should boost US manufacturing employment for vertically-integrated firms. The results are different for horizontal foreign investment, as reported in column (3) of Table 5. The coefficient on both high- and low-income affiliate wages is now positive and statistically significant. The coefficient on low-income affiliate wages, at 0.269, suggests that a 10 percent fall in affiliate wages would be associated with a 2.7 percentage point fall in parent employment. The coefficient on high-income affiliate wages is

also positive and statistically significant. These results suggest that affiliate employment in horizontally-integrated multinationals substitutes for parent employment.

The own-wage elasticity, which varies between -0.34 and -0.52, suggests that a one percent increase in the domestic US manufacturing wage reduces labor demand by 0.34 to 0.52 percent. The magnitude is in line with the dozens of studies cited in Hamermesh (1993), who reports that most studies find that the own-wage elasticity for labor lies between 0.3 and 0.7. The coefficient on the industry-specific home price of investment is positive for both types of enterprises, suggesting that reductions in the price of domestic investment goods reduce domestic labor demand. The coefficient on investment abroad has the opposite sign, suggesting that reductions in the cost of investment abroad are associated with increased labor demand at home. The coefficient estimates on the domestic price of investment goods imply that in the US, investment and labor are generally substitutes. This is consistent with a story in which less skilled workers are being replaced by capital (computers) and consistent with previous labor demand studies on capital-labor substitution cited in Hamermesh (1993).

Increases in import penetration (arms-length trade) and foreign demand shocks also affect domestic labor demand. A 10 percentage point increase in import penetration during the sample period would imply a decline in US manufacturing employment of 3.5 percentage points for vertical firms. For horizontal enterprises, import penetration from low-wage countries significantly and negatively affected labor demand. The point estimates imply that a 10 percentage point increase in import penetration from low-wage countries was associated with a 6.1 percentage point fall in labor demand. The different effects for horizontal and vertical firms across different types of import competition reflect the fact that there was significantly more import competition from low wage countries in the industries which we characterize as horizontal, including textiles and apparel, leather products, pulp and paper, and woods products. A one percent increase in the prices of consumption goods in affiliate locations is associated with a 0.58 percentage point increase in employment for vertical multinationals and a 1.15 increase for horizontal multinationals.

Big negative employment effects are also associated with our domestic proxy for the prices of technology inputs, the share of research and development employment in total parent employment. The results indicate that a ten percentage point increase in the parent research and development employment

shares would be associated with a 3.5 percentage point decline for vertical firms and 9.1 percentage point decline for horizontal firms. For horizontally-integrated firms, parent research and development employment as a share of total employment increased by 4.3 percentage points between 1977 and 1999, implying a reduction in home employment of 4 percent. However, the coefficients on affiliate R&D employment are either insignificant or positive, suggesting that R&D activities in affiliates are associated with positive employment gains for the US parent.

It is also interesting to note the different coefficients on the dummy variable for activities in low-income affiliate locations. The coefficient is positive and significant for vertical parents but imprecisely estimated and not significant from zero for horizontal parents. For vertical parents, the coefficient suggests that a one percentage point increase in low income affiliate employment would be associated with a .93 percentage point increase in parent employment. The dummy variable for low-income affiliate activity can be interpreted as a measure of the impact of changes in the extensive margin of multinational activity abroad, where the extensive margin reflects changes in affiliate locations while the intensive margin reflects increases in activity in existing locations (see, for example, Muendler and Becker (2006) for a discussion of extensive versus intensive margins for German and Swedish multinationals). The results suggest that for vertical enterprises, more offshoring activities via the intensive or extensive margins is associated with increasing labor demand for the US parent.

The critical parameters of interest in Table 5 are the coefficients on affiliate wages, which indicate whether affiliate employment substitutes for or is complementary with home employment. In Table 6, we explore whether our results are robust to the definition of affiliate wages. Instead of constructing country-level wages from the BEA sample, we use country wages reported by UNIDO. Wages are calculated based on surveys administered by UNIDO, supplemented with secondary sources (such as national statistical agencies) gathered by UNIDO as well. Wages are calculated as compensation divided by number of employees, collected at the 3 digit ISIC level (Revision 2). All values are converted to US dollars using the IMF exchange rate series *rf*. As in Table 5, we weight country-level wages using the parent's initial distribution of employment across affiliate locations when the parent first appears in the sample. The results in Table 6 are consistent with our earlier results, suggesting that the source for country-level wages does not affect our coefficient estimates. The coefficients on high- and

Comment [aw2]: You haven't described this dataset at all yet, do you think you need to?

low-income affiliate wages are the same sign and close in magnitude to the previous results. As before, the results indicate that home and foreign employment are complements for vertical multinationals but that home and foreign employment are substitutes for horizontal multinationals. The coefficient estimates on low income wages suggest that a 1 percent increase in foreign affiliate wages would lead to a .31 percent reduction in parent employment for horizontal firms and a .25 percent increase in employment for vertical firms.

Alternative Specifications: Selection, Translog and CES Specifications

We face potentially important selection problems. Between each benchmark year, roughly 20% of the parents drop out of our sample and do not reappear. If some of these firms relocate *all* operations abroad and close down their US operations, then our estimates of the employment costs of multinational activity could be downward biased. Following Wooldridge (2002, p. 581), we test for survivorship bias by including a lead of the selection indicator $s_{i,t+1}$ in our estimating equations, where $s_{i,t+1}$ is equal to zero for firms that do not exit the sample and switches from zero to one in the period just before attrition. The coefficient on the lead of the selection indicator was negative and significant for both vertically- and horizontally-integrated firms. The significant and negative sign on the selection variable is a possible indicator that firms which exit the sample are those most likely to contract employment. To address this potential criticism, we correct for selection bias using two approaches: a Heckman type selection correction and inverse probability weighting.

Following Wooldridge (2002) our first approach—a Heckman-type correction--models this selection problem as follows. If our equation of interest is given by:

$$y_{it} = x_{it}\beta + u_{it}, \quad t = 2, \dots, T,$$

then conditional on the parent reporting in the previous period, i.e. $s_{i,t-1} = 1$, we can write a reduced form selection equation for $t \geq 2$ as,

$$s_{it} = \mathbb{I}[w_{it}\delta_t + v_{it} > 0], \quad \text{where,} \quad v_{it} \mid \{x_{it}, w_{it}, s_{i,t-1} = 1\} \sim \text{Normal}(0,1)$$

In the context of panel data with an unobserved firm fixed effect and attrition, Wooldridge (2002) proposes as a solution a variant of a two-stage Heckman correction. In each period, Wooldridge proposes estimating a selection equation using a probit approach and calculating lambda, the inverse Mills ratio, for each parent i . Once a series of lambdas have been estimated for each year and parent, the estimating equations are augmented by these lambdas.

This approach is only successful if we can identify determinants of the binary selection variable s_{it} before the firm exits the sample (in period $t-1$) and which do not belong in the estimating equation. We identified candidate variables using the insights derived from a class of models indicating that heterogeneity in productivity is a significant determinant of whether firms enter into international trade or foreign investment (see Melitz (2003)). These models suggest that only the most profitable firms are likely to engage in trade or foreign investment. Since we already control for output and factor price shocks using a variety of input and output prices, parent profitability in the previous period does not belong in the estimating equations (indeed, auxiliary regressions show that lagged profits from the benchmark surveys five years earlier do not predict current period employment). Consequently, we use as the excluded determinant of survival the parent's profitability in the previous period.

Appendix Table A5 reports estimates using this two-step approach. The sample size decreases significantly, since implementing the selection correction eliminates the first time series observation for each parent. A joint F-test for the inclusion of the selection terms is statistically significant. The coefficient on the inverse Mills ratio is positive and significant for vertically integrated multinationals and negative and significant for horizontally integrated firms. The sign on the inverse Mills ratio indicates the nature of the correlation between the errors in the selection equation and the second stage equation of interest. Since the sign on lambda is positive for vertically integrated firms, this suggests that those firms most likely to survive are also most likely to have unanticipated positive draws for domestic employment. The negative coefficient for horizontally integrated firms suggests the opposite: firms most likely to survive are also more likely to have unanticipated lower employment.

Adding the inverse Mills ratio to control for selection does not change the sign and barely changes the point estimates on the coefficients of interest. The coefficients on affiliate wages remain negative and statistically significant for vertically integrated multinationals operating in low income

affiliates, suggesting that employment in these locations are complementary with parent employment. For horizontal firms, the coefficients on affiliate wages remain positive and of similar magnitude to our earlier results. Since we lose almost 50 percent of the sample when we correct for attrition, in the remainder of the paper we do not control for attrition bias.

We also explored the use of using inverse probability weighting as outlined in Wooldridge (2002) to correct for selection bias. This approach consists of the following two-step procedure. In each time period, we estimate a binary response model for the probability of survival for the group in the sample at time $t-1$. Using the fitted probabilities from the first step, we obtain the following weights:

$$\hat{p}_{it} = \hat{\pi}_{it} * \hat{\pi}_{i,t-1} * \dots * \hat{\pi}_{i,1}.$$

where hats denote fitted probabilities. This methodology allowed us to choose covariates in the probits that are essentially everything we can observe for units in the sample at time $t-1$ that might affect attrition. In our case, we included all of the regressors in our original model plus firm size, firm profitability and the firm's share of employment in low-income countries. Using this approach also did not affect the robustness of our estimates and consequently we do not report them here.

We also test for the robustness of our results to two alternatives: a framework based on a translog cost function and a framework based on CES production functions. The translog approach has been adopted by Brainard and Riker (2001), Hanson, Mataloni and Slaughter (2003) and Muendler and Becker (2006). This alternative approach has the advantage that the translog cost function approximates many well behaved cost functions. The translog total variable cost (TC) function (omitting time and parent subscripts) for wages W , investment prices r , research and development input prices t and output Y is given by:

$$\begin{aligned}
\ln TC = & \alpha_0 + \sum_j \varpi_j \ln Y + \sum_j \alpha_{jw} \ln W + \sum_j v_j \ln r + \sum_j \alpha_{jA} \ln t \\
& + \frac{1}{2} \sum_j \sum_k \alpha_{jY} (\ln Y)^2 + \frac{1}{2} \sum_j \sum_k \xi_{jk} (\ln W)^2 + \frac{1}{2} \sum_j \sum_k \beta_{jk} (\ln t)^2 \\
(5') \quad & + \frac{1}{2} \sum_j \sum_k \omega_{jk} (\ln r)^2 + \sum_j \sum_k \vartheta_{jk} \ln W \ln r + \sum_j \sum_k \tau_{jk} \ln Y \ln t \\
& + \sum_j \sum_k \rho_{jk} \ln Y \ln W + \sum_j \sum_k \chi_{jk} \ln r \ln t + \sum_j \sum_k \varphi_{jk} \ln r \ln Y + \sum_j \sum_k \kappa_{jk} \ln t \ln W + \varepsilon
\end{aligned}$$

Differentiating $\ln TC$ with respect to $\ln W_j$ according to Shepard's lemma, and allowing for a firm fixed effect, yields labor's share in total costs in location j for parent i at time t :

$$(6) \quad LSHARE_{ijt} = \beta_0 + \sum_j \rho_j \ln Y_{ijt} + \sum_j \kappa_j \ln t_{ijt} + \sum_j \xi_j \ln w_{ijt} + \sum_j \vartheta_j \ln r_{ijt} + f_i + \varepsilon_{ijt},$$

where $LSHARE$ is defined as the cost share of labor expenditures in location j for parent i in time t , relative to expenditures on labor and capital across all locations. We impose the restrictions implied by the framework; in particular, it must be the case that the coefficients on factor prices sum to zero.

For completeness, we also consider aggregating capital and labor across locations using a CES function (Katz and Murphy (1992) and Card (2001) use this approach). Thus we define L as follows:

$$(7) \quad L_i = \left[\sum_j (e_{ij} N_{ij})^{\frac{\sigma-1}{\sigma}} \right]^{\frac{\sigma}{\sigma-1}}$$

where e represents productivity shocks, L_i is the total quantity of labor used, and σ is the Allen elasticity of substitution between labor in location i and j and is defined below.¹⁰

The first-order condition with respect to labor hired in the US is:

$$(8) \quad p_{i,h} Y_L \frac{\partial L_i}{\partial L_{i,h}} = w_{i,h}$$

The first-order condition with respect to labor hired in high-income affiliates is:

¹⁰ If sigma is equal to zero, we have the case of perfect complements (i.e. left shoes and right shoes, the leontief function that looks like $L = \min(L_h, L_f)$) this is obviously extreme but might be applicable to some kinds of natural resource extraction. The polar opposite is σ tending to infinity (i.e. labor at home and labor abroad are perfect substitutes so $L = L_h + L_f$) – this is also extreme but some version of this might be realistic for production workers.

$$(9) \quad p_{i,hif} Y_L \frac{\partial L_i}{\partial L_{i,hif}} = w_{i,hif}$$

where p are final goods prices at home and abroad and w are wages at home and abroad. The first-order condition with respect to labor hired in low-income affiliates is:

$$(10) \quad p_{i,lif} Y_L \frac{\partial L_i}{\partial L_{i,lif}} = w_{i,lif}$$

Since,

$$(11) \quad \frac{\partial L_i}{\partial L_{i,h}} = \frac{\sigma}{\sigma-1} \left[\sum_j (e_{ij} L_{ij})^{\frac{\sigma-1}{\sigma}} \right]^{\frac{1}{\sigma-1}} \frac{\sigma-1}{\sigma} \left[(e_{i,h} L_{i,h}) \right]^{\frac{-1}{\sigma}} e_{i,h}$$

and,

$$(12a) \quad \frac{\partial L_i}{\partial L_{i,hif}} = \frac{\sigma}{\sigma-1} \left[\sum_j (e_{ij} L_{ij})^{\frac{\sigma-1}{\sigma}} \right]^{\frac{1}{\sigma-1}} \frac{\sigma-1}{\sigma} \left[(e_{i,hif} L_{i,hif}) \right]^{\frac{-1}{\sigma}} e_{i,hif}$$

$$(12b) \quad \frac{\partial L_i}{\partial L_{i,lif}} = \frac{\sigma}{\sigma-1} \left[\sum_j (e_{ij} L_{ij})^{\frac{\sigma-1}{\sigma}} \right]^{\frac{1}{\sigma-1}} \frac{\sigma-1}{\sigma} \left[(e_{i,lif} L_{i,lif}) \right]^{\frac{-1}{\sigma}} e_{i,lif}$$

we can insert (11) and (12) into (8) through (10) and take the following:

$$(13a) \quad \frac{p_{ih} e_h^{\frac{\sigma-1}{\sigma}} L_h^{\frac{-1}{\sigma}}}{p_{i,hif} e_{i,hif}^{\frac{\sigma-1}{\sigma}} L_{i,hif}^{\frac{-1}{\sigma}}} = \frac{w_{ih}}{w_{i,hif}}$$

$$(13b) \quad \frac{p_{ih} e_h^{\frac{\sigma-1}{\sigma}} L_h^{\frac{-1}{\sigma}}}{p_{i,lif} e_{i,lif}^{\frac{\sigma-1}{\sigma}} L_{i,lif}^{\frac{-1}{\sigma}}} = \frac{w_{ih}}{w_{i,lif}}$$

Taking logs of both sides of (13a) and (13b) yields the following:

$$(14a) \quad \ln(L_h / L_{hif}) = \sigma \ln \frac{p_h}{p_{hif}} + (\sigma-1) \ln \frac{e_h}{e_{hif}} - \sigma \ln \frac{w_h}{w_{hif}} .$$

$$(14b) \quad \ln(L_h / L_{lif}) = \sigma \ln \frac{p_h}{p_{lif}} + (\sigma-1) \ln \frac{e_h}{e_{lif}} - \sigma \ln \frac{w_h}{w_{lif}} .$$

Equations (14a) and (14b) underscore the fact that as long as there is some substitution (or complementarity) between domestic and foreign labor, the cost of labor abroad plays an important role in determining the demand for US labor. In addition, one of the restrictions of the CES specification is that

the Allen elasticity of substitution between parent and low-income affiliates should be the same as the elasticity of substitution between parent and high-income affiliates.

Comparing Elasticities of Labor Demand Across Specifications

All three approaches yield coefficient estimates which can be used to derive elasticities of factor demand η and Allen elasticities of substitution σ . In equation (5), the key parameters are the elasticities of factor demand η . Typically, inputs i and j are referred to as p-complements if η_{ij} is less than zero, and p-substitutes if η_{ij} is greater than zero. The key parameters in equation (6) are the ξ_j 's. To convert these into Allen partial elasticities of substitution between locations, we can calculate the following based on observed labor shares s_j :

$$(15) \sigma_{jk} = (\xi_{jk} + s_j s_k) / s_j s_k$$

$$\sigma_{jj} = (\xi_{jj} + s_j s_j - s_j) / s_j s_j$$

The Allen partial elasticity of substitution σ_{jk} gives us the percentage change in the ratio of L_j to L_k with respect to the percentage change in the ratio of w_k to w_j . The Allen partial elasticity of substitution is directly estimated as the coefficient on relative wages using the CES approach (equations (14a) and (14b)). To convert the Allen partial elasticity of substitution into an elasticity of factor demand, we multiply by the factor share:

$$(16) \eta_{ij} = s_j \sigma_{ij} = \partial \ln L_i / \partial \ln w_j$$

We report elasticities of substitution for each of the three estimation strategies in Table 7. Factor shares are typically computed by taking the sample means of the data.¹¹ For the translog approach, we report the implied elasticities from estimating equation (6), substituting our price measures for output Y to avoid endogeneity concerns, in Table 7. The coefficients on affiliate wages imply that foreign labor in

¹¹ Confidence intervals could be computed using bootstrapped standard errors.

horizontal multinationals substitutes for home labor in both high- and low-income affiliate locations. For vertical multinationals, the results are the opposite: workers in low-income locations are complementary to domestic employees. As expected, the own-price elasticity is negative. The results are generally consistent with our labor demand specification reported in Tables 5 and 6.

For vertical multinationals, the point estimates are consistently negative in low-income locations but not precisely estimated for high-income locations. The results imply that low-income employment is complementary with domestic employment in vertical multinationals. The CES approach corroborates the results in Tables 5 and 6 showing that affiliate labor in horizontal multinationals substitutes for parent employment. For vertical multinationals, labor in low-income affiliates is complementary with US parent employment.

We summarize the effects of factor price changes, trade, and technical change on US manufacturing employment in Table 8. We combine the coefficient estimates presented in Table 5 with the actual mean changes in wages, investment prices, trade, research and development employment, and goods prices taken from Table 3. We see that the major determinants of contraction in US manufacturing parent employment are (1) falling real prices of consumption goods (2), falling prices of investment goods (which incorporate the falling prices of computers) (3) falling real wages in low-income affiliate locations and (4) increasing import competition. While much of the debate on offshoring focuses in (3), the impact of relative wage changes on US parent labor demand is only one factor that explains contraction in parent employment. For horizontal multinationals, the combined effects of higher domestic wages and falling foreign wages only account for a 5.5 percent decline in US employment. In comparison, falling consumption prices account for an 11 percent decline, falling investment prices at home (leading firms to substitute capital for labor) account for a 7.5 percent decline, and increasing import competition from low-wage countries accounts for a 6 percent decline in home employment.

For vertical multinationals, falling real wages in low-income affiliates *boosted* employment. Real declines in affiliate wages account for an 8.5 percent increase in parent labor demand. However, falling domestic investment prices and increasing import competition offset this gain, leading to a net contraction in parent employment. Even for vertical firms, the gains from operating in low-income countries were not sufficient to offset the negative effects on domestic parent employment of falling domestic investment

prices, increasing manufacturing wages at home, falling consumer prices abroad, and increasing import competition.

We have not allowed for the possibility that hiring workers abroad prevents a parent from shutting down operations. Borjas, Freeman and Katz (1997) study this counterfactual by asking what would have happened to low wage workers if imports from developing countries had been produced by US firms. In a similar spirit, we would like to know what might have happened to US workers if the parent had not hired workers in developing countries. Our dummy for employment in developing countries suggests that low-income affiliate activities have, in fact, preserved some jobs. Table 8 shows that firms which expanded operations in low-income affiliate locations on average increased parent employment by 1.6 (for horizontal) to 2.6 (for vertical) percentage points. For horizontal firms, the coefficient is not generally significant. While the point estimate is large in magnitude and significant for vertical firms, the net impact is not large because the change in the extensive margin (ie the increase in the percentage of firms operating in low-income countries) over the sample period is quite small. What this suggests is that for US multinationals, changes in the intensive margin in the past were significantly more important in explaining US employment changes than movements in the extensive margin. One issue for future research is the potential endogeneity of the dummy variable for employment in developing countries. Given that larger and more productive firms are more likely to have operations in low income affiliate locations, the positive impact at the intensive margin probably overstates the domestic employment gains for vertical multinationals.

V. Concluding Comments

This paper measures the impact of different forms of globalization on manufacturing employment by US multinationals in the United States. Over the period 1982 to 1999 domestic employment of US multinationals contracted by nearly 4 million jobs, possibly foreshadowing the overall reduction in US manufacturing employment that accelerated from 1999 onwards. During this period, the number of workers hired by affiliates in developing countries increased while wages paid to these workers declined.

These facts are consistent with the hypothesis that US parents are exporting low-wage jobs to low-income countries. In this paper, we show that this hypothesis is only partly supported by the evidence.

The expansion of manufacturing employment in developing countries amounts to only one-quarter of the jobs lost at home. Our research shows that other factors—including technological change, falling domestic investment goods prices, and import competition—are more important determinants of falling US manufacturing employment. Moreover, we find that for vertical multinationals, increasing offshore activities appears to have preserved jobs, rather than eliminated them. These results are consistent with research by Grossman and Rossi-Hansberg (2006) suggesting that vertical activities could act as a positive technology shock, increasing domestic labor demand. This is the first paper to provide empirical support for this theoretical possibility.

Using data on US based multinationals from the Bureau of Economic Analysis (BEA), we measure the impact on US manufacturing employment of changes in foreign affiliate wages, controlling for changing demand conditions, import competition and technological change. We find that the evidence on the links between offshoring and domestic employment is mixed, and that the effect depends on both the type and the location of foreign investment. We conclude that the heterogeneity in effects is one reason why previous research on this topic has yielded such apparently contradictory results. For firms most likely to perform the same tasks in foreign affiliates and at home (“horizontal” foreign investment), foreign and domestic employees are substitutes. For these firms, lower wages in affiliate locations are associated with lower employment in the US: a 1 percentage point fall in affiliate wages is associated with a .27 percent fall in parent employment. However, for firms which do significantly different tasks at home and abroad (“vertical” foreign investment), foreign and domestic employment are complements: a 1 percentage point decline in low-income affiliate wages is associated with a .32 percent increase in parent employment. The complementarity between domestic and foreign activities for vertically-integrated firms is consistent with theoretical models developed by Markusen, Maskus, Grossman and Rossi-Hansberg.

For both vertical and horizontal parents, other factors play an important role in explaining the contraction in US manufacturing employment. For horizontally integrated multinationals firms, falling consumption prices account for an 11 percent decline, falling investment prices which have allowed firms

to replace workers with computers or machines account for a 9 percent decline, and increasing import competition from low-wage countries accounts for a 6 percent decline in home employment. For vertical multinationals, the gains from operating in low-income countries were not sufficient to offset the negative effects on domestic parent employment of falling investment prices, falling consumer prices abroad, and increasing import competition.

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TABLE 1
TRENDS IN MANUFACTURING EMPLOYMENT BY US MULTINATIONALS 1982-1999

| | | 1982 | 1989 | 1994 | 1999 |
|--------------------------------------|------------|--------|--------|-------|-------|
| Parents | | | | | |
| (1) | BEA Mfg | 11,758 | 10,706 | 9,622 | 7,954 |
| (2) | Our sample | 10,689 | 9,668 | 9,104 | 7,564 |
| | Vertical | 5,812 | 5,581 | 4,958 | 4,263 |
| | Horizontal | 4,877 | 4,087 | 4,146 | 3,301 |
| High-income Affiliates | | | | | |
| | Total | 2,595 | 3,171 | 3,048 | 2,903 |
| | Vertical | 1,664 | 2,124 | 1,901 | 1,905 |
| | Horizontal | 932 | 1,048 | 1,147 | 998 |
| Low-income Affiliates | | | | | |
| | Total | 1,064 | 1,405 | 1,584 | 1,868 |
| | Vertical | 589 | 833 | 894 | 1,105 |
| | Horizontal | 475 | 571 | 690 | 764 |
| All Affiliates | | | | | |
| | Total | 3,659 | 4,576 | 4,632 | 4,772 |
| | Vertical | 2,253 | 2,957 | 2,795 | 3,010 |
| | Horizontal | 1,406 | 1,619 | 1,837 | 1,762 |
| Affiliate Share of Employment | | | | | |
| | Total | 26% | 32% | 34% | 39% |
| | Vertical | 28% | 35% | 36% | 41% |
| | Horizontal | 22% | 28% | 31% | 35% |

(1) Source: Mataloni (1994) and Mataloni and Yorgason (2006), Employment is by Industry of Parent and includes petroleum extraction and refining.

(2) Our totals differ from the BEA's because we drop observations for which wages, employment, R&D spending and/or R&D employment are negative.

TABLE 2:
DEFINING HORIZONTAL AND VERTICAL FOREIGN DIRECT INVESTMENT

| Industry (97 NAICS code) | Intrafirm Dummy | Mean Intrafirm Trade | Tariffs | | |
|-----------------------------|--------------------|----------------------------|--|--|----------------------------------|
| | | | USA (Bernard, Jensen, and Schott (2006)) 1979 | Developing Countries (World Bank) 1999 | China (World Bank) 1999 |
| Textiles and Apparel | -0.004 [0.004] | .008 | 26.3 | 14.6 | 26.3 |
| Food | -0.003 [0.004] | .018 | 15.2 | 14.3 | 33.6 |
| Beverages and Tobacco | 0.004 [0.005] | .016 | 27.3 | 29.4 | 63.8 |
| Leather Products | -0.005 [0.010] | .019 | 16.0 | 10.2 | 12.6 |
| Wood Products | 0.053 [0.514] | .013 | 14.0 | 8.1 | 7.2 |
| Paper | -0.006 [0.004] | .016 | 9.9 | 7.5 | 10.7 |
| Petroleum and Coal Products | 0.003 [0.005] | .027 | 5.1 | 6.2 | 9.2 |
| Chemicals | 0.039 [0.004]** | .072 | 9.7 | 7.0 | 14.0 |
| Plastics and Rubber | 0.016 [0.004]** | .051 | 16.2 | 10.3 | 14.6 |
| Nonmetallic Minerals | 0.002 [0.005] | .024 | 17.3 | 9.5 | 17.1 |
| Primary Metals | 0.010 [0.005]* | .043 | 12.1 | 6.8 | 8.4 |
| Fabricated Metals | 0.017 [0.004]** | .048 | 14.0 | 8.9 | 11.4 |
| Machinery | 0.049 [0.004]** | .082 | 9.0 | 6.2 | 13.0 |
| Computer and Electronics | 0.110 [0.005]** | .143 | 7.9 | 5.8 | 12.8 |
| Electrical Equipment | 0.042 [0.006]** | .074 | 10.8 | 7.5 | 12.2 |
| Transportation Equipment | 0.021 [0.004]** | .055 | 10.0 | 10.3 | 16.8 |
| Furniture | 0.011 [0.007] | .014 | 16.0 | 10.8 | 22.1 |
| Miscellaneous | 0.060 [0.005]** | .093 | 12.5 | 10.5 | 16.5 |
| Observations | 3866 | | | | |
| R-squared | 0.25 | | | | |

Notes: Robust standard errors in brackets, * significant at 5%; ** significant at 1%

For a detailed description of which industries are included in miscellaneous see:

<http://www.census.gov/epcd/naics/NAICS33C.HTM#N339>. Intrafirm trade is defined as the sum of exports to foreign affiliates for further processing plus imports from foreign affiliates, divided by sales.

TABLE 3
CLASSIFICATION OF COUNTRIES INTO LOW VERSUS HIGH INCOME CATEGORIES

| Country | World Bank / Our Classification for Low versus High Income | Nominal Manufacturing Wages (1994) in US Dollars From BEA Database |
|---------------------|--|---|
| Estonia | Low | 1,470 |
| Guyana | Low | 1,504 |
| China | Low | 1,579 |
| Malawi | Low | 1,689 |
| Romania | Low | 1,866 |
| Sri Lanka | Low | 1,898 |
| Ukraine | Low | 2,151 |
| India | Low | 2,325 |
| Dominican Republic | Low | 2,763 |
| Tanzania | Low | 3,057 |
| Zimbabwe | Low | 3,109 |
| Uzbekistan | Low | 3,136 |
| Zambia | Low | 3,152 |
| Vietnam | Low | 3,326 |
| Indonesia | Low | 3,401 |
| Botswana | Low | 3,517 |
| Pakistan | Low | 3,631 |
| Nigeria | Low | 3,940 |
| Honduras | Low | 4,111 |
| Thailand | Low | 4,168 |
| Costa Rica | Low | 4,236 |
| Yemen, Rep. | Low | 4,248 |
| Senegal | Low | 4,318 |
| Philippines | Low | 4,427 |
| Slovak R. | Low | 4,531 |
| Colombia | Low | 4,603 |
| El Salvador | Low | 4,622 |
| Egypt, Arab Rep. | Low | 4,756 |
| Fiji | Low | 4,824 |
| Kenya | Low | 5,098 |
| Malaysia | Low | 5,334 |
| Hungary | Low | 5,426 |
| Ghana | Low | 5,475 |
| Poland | Low | 5,540 |
| Jamaica | Low | 5,557 |
| Ecuador | Low | 5,596 |
| Panama | Low | 6,453 |
| Mexico | Low | 6,465 |
| Guatemala | Low | 6,786 |
| Trinidad and Tobago | Low | 6,994 |
| Venezuela, RB | Low | 7,393 |
| Swaziland | Low | 7,500 |
| Russian Federation | Low | 7,527 |
| Uruguay | Low | 7,997 |
| Turkey | Low | 8,370 |

| | | |
|----------------------|------|--------|
| Morocco | Low | 8,422 |
| Tunisia | Low | 9,058 |
| Nicaragua | Low | 9,206 |
| Malta | Low | 9,211 |
| Chile | Low | 9,485 |
| South Africa | Low | 10,257 |
| Barbados | Low | 10,480 |
| Peru | Low | 11,065 |
| Brazil | Low | 11,227 |
| Singapore | High | 11,885 |
| Portugal | High | 14,236 |
| Bahamas, The | High | 14,288 |
| Taiwan | High | 14,699 |
| Saudi Arabia | Low | 14,912 |
| Korea, Rep. | High | 15,549 |
| Bahrain | High | 16,047 |
| Netherlands Antilles | High | 16,596 |
| Hong Kong, China | High | 17,478 |
| New Zealand | High | 17,736 |
| Argentina | Low | 18,003 |
| Israel | High | 19,572 |
| Greece | High | 22,855 |
| Australia | High | 23,313 |
| Ireland | High | 23,392 |
| Spain | High | 25,848 |
| United Kingdom | High | 26,487 |
| Sweden | High | 27,380 |
| Italy | High | 30,574 |
| Austria | High | 31,209 |
| Finland | High | 32,049 |
| Denmark | High | 32,934 |
| Norway | High | 33,022 |
| United Arab Emirates | High | 33,603 |
| France | High | 33,628 |
| Aruba | High | 34,745 |
| Canada | High | 35,268 |
| Netherlands | High | 35,973 |
| Belgium | High | 40,134 |
| Luxembourg | High | 43,614 |
| Germany | High | 44,146 |
| Switzerland | High | 44,248 |
| Japan | High | 57,126 |

TABLE 4
SUMMARY STATISTICS: VERTICALLY-INTEGRATED FIRMS

| Variable | No. of Obs | Mean | Standard Deviation | Change in 1982- 1999 |
|--|---------------|--------|-----------------------|----------------------------|
| Log US employment | 2088 | 7.327 | 1.570 | -0.369 |
| US (domestic) Share in Labor Expenditures across all locations | 2088 | 0.807 | 0.131 | -0.041 |
| High-Income Affiliate Share in Labor Expenditures | 2088 | 0.172 | 0.116 | 0.038 |
| Low-Income Affiliate Share in Labor Expenditures | 2088 | 0.021 | 0.040 | 0.003 |
| Log US Manufacturing Wages, BEA | 2088 | 3.394 | 0.155 | 0.166 |
| Log High-Income Affiliate Wages | 2088 | 2.979 | 0.380 | -0.170 |
| Log Low-Income Affiliate Wages | 2088 | 2.319 | 1.088 | -0.266 |
| Log US Price of Investment, NBER Manufacturing Database | 2088 | 0.798 | 0.092 | -0.271 |
| Log Foreign Price of Investment, Penn World Tables (PWT) | 2088 | 0.675 | 0.264 | -0.079 |
| Log Foreign Price of Consumer Goods, PWT | 2088 | 0.712 | 0.283 | -0.059 |
| U.S R&D Employees (% in Total Employment) | 2088 | 0.091 | 0.091 | 0.043 |
| High-Income Affiliate R&D Employment (% in Total Employment) | 2088 | 0.036 | 0.063 | -0.024 |
| Low-Income Affiliate R&D Employment (% in Total Employment) | 2088 | 0.005 | 0.016 | -0.006 |
| Import Penetration, Bernard, Jensen, and Schott (2006) | 2088 | 0.178 | 0.108 | 0.146 |
| Import Penetration from Low-Income Countries, Bernard et al (2006) | 2088 | 0.044 | 0.040 | 0.062 |
| Percent Firms with Employees in Low-income Countries | 2088 | 0.359 | 0.499 | 0.028 |
| Parent Sales by Industry | 2088 | 12.225 | 1.564 | 0.096 |
| Affiliate Sales by Industry | 2088 | 9.649 | 1.406 | 0.348 |

SUMMARY STATISTICS: HORIZONTALLY-INTEGRATED FIRMS

| Variable | No. of Obs | Mean | Standard Deviation | Change in 1982- 1999 |
|--|---------------|--------|-----------------------|----------------------------|
| Log US employment | 1778 | 7.928 | 1.775 | -0.239 |
| US (domestic) Share in Labor Expenditures across all locations | 1778 | 0.834 | 0.137 | -0.039 |
| High-Income Affiliate Share in Labor Expenditures | 1778 | 0.140 | 0.126 | 0.036 |
| Low-Income Affiliate Share in Labor Expenditures | 1778 | 0.025 | 0.046 | 0.005 |
| Log US Manufacturing Wages, BEA | 1778 | 3.322 | 0.204 | 0.009 |
| Log High-Income Affiliate Wages | 1778 | 3.001 | 0.363 | -0.199 |
| Log Low-Income Affiliate Wages | 1778 | 2.208 | 1.080 | -0.189 |
| Log US Price of Investment, NBER Manufacturing Database | 1778 | 0.827 | 0.040 | -0.131 |
| Log Foreign Price of Investment, Penn World Tables (PWT) | 1778 | 0.647 | 0.259 | -0.117 |
| Log Foreign Price of Consumer Goods, PWT | 1778 | 0.674 | 0.276 | -0.094 |
| U.S R&D Employment (% in Total Employment) | 1778 | 0.031 | 0.050 | 0.003 |
| High-Income Affiliate R&D Employment (% in Total Employment) | 1778 | 0.020 | 0.058 | -0.023 |
| Low-Income Affiliate R&D Employment (% in Total Employment) | 1778 | 0.006 | 0.023 | -0.009 |
| Import Penetration, Bernard, Jensen, and Schott (2006) | 1778 | 0.127 | 0.085 | 0.082 |
| Import Penetration from Low-Income Countries, Bernard et al (2006) | 1778 | 0.053 | 0.059 | 0.099 |
| Percent Firms with Employees in Low-income Countries | 1778 | 0.442 | 0.476 | 0.094 |
| Parent Sales by Industry | 1778 | 12.828 | 1.733 | -0.082 |
| Affiliate Sales by Industry | 1778 | 9.746 | 1.704 | 0.212 |

Unless indicated, variables are computed using the BEA benchmark surveys of direct investment abroad for the years 1982, 1989, 1994, 1999.

TABLE 5
WITHIN ESTIMATES OF LABOR DEMAND: US PARENTS

| | (1) | (2) | (3) | (4) |
|--|---------------------|---------------------|---------------------|--|
| | Pooled | Vertical | Horizontal | F-test (p-value) for equality of coefficients) |
| Log US Industrial Wages | -0.390 [0.070]** | -0.336 [0.080]** | -0.518 [0.090]** | 0.105 |
| Log Industrial Wages in Low Income Countries | 0.067 [0.033]* | -0.318 [0.065]** | 0.269 [0.084]** | 0.000 |
| Log Industrial Wages High Income Countries | -0.031 [0.015]* | -0.060 [0.291] | 0.092 [0.036]** | 0.034 |
| Log of the US Price of Capital | 0.251 [0.160] | 0.474 [0.181]** | 0.573 [0.406] | 0.834 |
| Log of the Foreign Price Of Capital | -0.338 [0.150]* | -0.121 [0.205] | -0.587 [0.212]** | 0.114 |
| Log of the Foreign Price Of Consumer Goods | 0.816 [0.149]** | 0.577 [0.198]** | 1.150 [0.208]** | 0.032 |
| Import Penetration | 0.027 [0.120] | -0.350 [0.116]** | 0.531 [0.491] | 0.000 |
| Import Penetration from Low Wage Countries | -0.138 [0.204] | 0.327 [0.291] | -0.610 [0.274]** | 0.000 |
| R&D (% Employment) | -0.475 [0.100]** | -0.354 [0.115]** | -0.913 [0.194]** | 0.013 |
| R&D (% Employment) In High-income Countries | 0.346 [0.164]* | 0.247 [0.190] | 0.723 [0.312]* | 0.014 |
| R&D (% Employment) in Low-income Countries | 1.399 [0.671]* | 1.905 [0.947]* | 1.271 [0.981] | 0.122 |
| Dummy Equal to One if Firm has Employees in Low-income Countries | 0.453 [0.097]** | 0.931 [0.143]** | 0.173 [0.128] | 0.000 |
| Time dummy 1989 | -0.023 [0.026] | 0.017 [0.036] | -0.001 [0.044] | 0.819 |
| Time dummy 1994 | -0.001 [0.036] | 0.105 [0.051]* | 0.002 [0.066] | 0.564 |
| Time dummy 1999 | 0.154 [0.046]** | 0.268 [0.064]** | 0.196 [0.090]* | 0.744 |
| Observations | 3866 | | 3866 | |
| Number of firms | 1868 | | 1868 | |
| R-squared | 0.05 | | 0.16 | |

Notes: Standard errors corrected for arbitrary heteroskedasticity are in brackets. * indicates significant at 5% while ** indicates significant at 1%. In column (1), coefficients on horizontal and vertical multinationals are the same. In columns (2) and (3) we allow the coefficients to differ. Column (4) reports tests of the equality of coefficients across the two types of enterprises. Dependent variable is the log of US employment. Within estimates calculated by taking deviations from firm-level means over the sample period.

TABLE 6
WITHIN ESTIMATES OF US LABOR DEMAND USING UNIDO WAGES INSTEAD OF
BEA WAGES IN LOW AND HIGH INCOME AFFILIATE LOCATIONS

| | (1) | (2) | (3) |
|--|---------------------|---------------------|---------------------|
| | Pooled | Vertical | Horizontal |
| Log US Industrial Wages | -0.311 [0.079]** | -0.299 [0.077]** | -0.399 [0.088]** |
| Log Industrial Wages in Low-income Countries | 0.076 [0.033]* | -0.252 [0.065]** | 0.311 [0.074]** |
| Log Industrial Wages in High-income Countries | -0.041 [0.023]* | -0.061 [0.029]* | 0.092 [0.036]** |
| Log of the US Price of Capital | 0.121 [0.060]* | 0.411 [0.112]** | 0.373 [0.226] |
| Log of the Foreign Price Of Capital | -0.338 [0.150]* | -0.111 [0.191] | -0.518 [0.209]** |
| Log of the Foreign Price Of Consumer Goods | 0.723 [0.149]** | 0.579 [0.198]** | 0.970 [0.333]** |
| Import Penetration | -0.007 [0.120] | -0.389 [0.155]** | 0.444 [0.694] |
| Import Penetration from Low Wage Countries | -0.267 [0.105]* | 0.307 [0.292] | -0.734 [0.334]** |
| R&D (% Employment) | -0.482 [0.100]** | -0.358 [0.115]** | -0.915 [0.194]** |
| R&D (% Employment) in High-income Countries | 0.345 [0.264] | 0.243 [0.190] | 0.720 [0.312]* |
| R&D (% Employment) in Low-income Countries | 1.356 [0.671] | 1.879 [0.999] | 1.259 [0.982] |
| Dummy equal to one if Firm has Employees in Low-income Countries | 0.332 [0.077]** | 0.734 [0.314]** | 0.183 [0.128] |
| Time dummy 1989 | -0.059 [0.038] | -0.073 [0.043] | -0.098 [0.101] |
| Time dummy 1994 | -0.062 [0.040] | -0.044 [0.045] | 0.057 [0.115] |
| Time dummy 1999 | 0.075 [0.046] | 0.082 [0.071] | 0.072 [0.071] |
| Observations | 3166 | 3166 | 3166 |
| Number of firms | 1529 | 1529 | |
| R-squared | 0.04 | 0.15 | |

Notes: Standard errors corrected for arbitrary heteroskedasticity are in brackets. * indicates significant at 5% while** indicates significant at 1%. In column (1), we impose that the coefficients on horizontal and vertical multinationals are the same. In columns (2) and (3) we allow the coefficients to differ. Dependent variable is the log of US employment. Within estimates calculated by taking deviations from firm-level means over the sample period. Log industrial wages in high and low income countries taken from UNIDO database, INDSTAT3 (2006), based on surveys and secondary sources collected by UNIDO and converted to US dollars.

TABLE 7
IMPLIED ELASTICITY OF LABOR DEMAND ACROSS ALTERNATIVE SPECIFICATIONS

| | (1) | (2) | (3) | (4) | (5) |
|---|------------------------|---|----------------------|---------------------------|---|
| | Basic Specification | Replacing BEA wages with UNIDO wages | CES specification | Translog Cost Function | Adding Sales to Basic Specification |
| | (Table 5) | (Table 6) | | | (Appendix Table A.4) |
| Implied Elasticity of Labor Demand η_{ij} (% Change in L_i in Response to % Change in w_j) | | | | | |
| Own Elasticity of Labor Demand | | | | | |
| Vertical | -0.336 | -0.299 | -- | -0.126 | -0.501 |
| Horizontal | -0.518 | -0.399 | -- | -0.153 | -0.658 |
| Elasticity with respect to wages in Horizontal Firms | | | | | |
| Low-income Affiliate Locations | 0.269 | 0.311 | 0.290 | 0.200 | 0.273 |
| High-income Affiliate Locations | 0.092 | 0.092 | 0.087 | 0.125 | 0.077 |
| Elasticity with respect to wages in Vertical Firms | | | | | |
| Low-income Affiliate Locations | -0.318 | -0.252 | -0.330 | -0.100 | -0.315 |
| High-income Affiliate Locations | -0.060 | -0.061 | 0.051 | 0.162 | -0.001 |

Notes: Coefficients taken from Tables 5 and 6 and unreported coefficients for CES and translog estimates. Factor shares used to compute elasticities taken from sample means (see Table 4).

Table 8
Calculating the Impact of Different Aspects of Globalization on Labor Market Outcomes

| Factors Affecting US Labor Demand | VERTICAL | | | | HORIZONTAL | | | |
|--|---------------------------------|---------------------------|-----------------------------------|---------------------------------------|---------------------------------|---------------------------|-----------------------------------|---------------------------------------|
| | Impact of 1% increase in factor | Actual increase in sample | Percentage Change in Labor Demand | Keeping Only Significant coefficients | Impact of 1% increase in factor | Actual increase in sample | Percentage Change in Labor Demand | Keeping Only Significant coefficients |
| | (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) |
| Log US Industrial Wages | -0.336 | 0.166 | -5.578 | -5.578 | -0.518 | 0.009 | -0.466 | -0.466 |
| Log Industrial Wages in Low-income Countries | -0.318 | -0.266 | 8.459 | 8.459 | 0.269 | -0.189 | -5.084 | -5.084 |
| Log Industrial Wages in High-income Countries | -0.060 | -0.170 | 1.020 | | 0.092 | -0.199 | -1.831 | -1.831 |
| Log of US Price of Capital | 0.474 | -0.271 | -12.845 | -12.845 | 0.573 | -0.131 | -7.506 | -7.506 |
| Log of Foreign Price of Capital | -0.121 | -0.079 | 0.956 | | -0.587 | -0.117 | 6.868 | 6.868 |
| Log of Foreign Price of Consumer Goods | 0.577 | -0.059 | -3.404 | -3.404 | 1.150 | -0.094 | -10.810 | -10.810 |
| Import Penetration | -0.350 | 0.146 | -5.125 | -5.125 | 0.531 | 0.082 | 4.354 | |
| Import Penetration from Low Wage Countries | 0.327 | 0.062 | 2.027 | | -0.610 | 0.099 | -6.039 | -6.039 |
| R&D Employment | -0.354 | 0.043 | -1.522 | -1.522 | -0.913 | 0.003 | -0.274 | -0.274 |
| R&D Employment in High-income Countries | 0.247 | -0.024 | -0.593 | | 0.723 | -0.023 | -1.663 | -1.663 |
| R&D Employment in Low-income Countries | 1.905 | -0.006 | -1.143 | -1.143 | 1.271 | -0.009 | -0.763 | |
| Dummy =1 if firm has employees in Developing Countries | 0.931 | 0.028 | 2.607 | 2.607 | 0.173 | 0.094 | 1.626 | |
| Net Impact of all Above Variables | | | -15.141 | -18.551 | | | -21.588 | -26.805 |

Notes: coefficients in columns (1) and (5) taken from Table 5. Numbers in columns (2) and (6) taken from means Table 4. Numbers in columns (3) and (7) calculated by multiplying by 100 x column (1) and column (2) (for column (3)) and column (5) and column (6) (for column (7)). Columns (4) and (8) calculated the same way as columns (3) and (7), but only the coefficients which were significant in Table 5 are reported. The Net Impact column sums up all the previous effects.

APPENDIX TABLE A.1: COVERAGE OF THE BEA SAMPLE

| Year and Variable | Coverage of BEA Sample in Manufacturing | Coverage of BEA Sample in Services | Coverage of BEA Sample in Total US Economic Activity (Includes Manufacturing, Services, Other, Wholesale Trade) |
|---|---|------------------------------------|---|
| 1982 | | | |
| Total Number of Employees in BEA Sample (Thousands) | 11,758.1 | 993.8 | 18,704.6 |
| Gross Product in the BEA Sample (US Millions of Dollars) | 421,050 | 25,997 | 796,017 |
| Coverage of the BEA Sample (in %) Relative to Gross Product for All Firms operating in the US | 80 % | 6 % | 33 % |
| Value of Dollar Export Sales by Firms in the BEA Sample (Millions) | 163,383 | NA | NA |
| Coverage of the BeA Sample (in %) Relative to Exports of All Firms operating in the US | 77 % | NA | NA |
| 1989 | | | |
| Total Number of Employees in BEA Sample (Thousands) | 10,706.8 | 1,700 | 18,785.4 |
| Gross Product in the BEA Sample (US Millions of Dollars) | 586,568 | 57,090 | 1,044,884 |
| Coverage of the BEA Sample (in %) Relative to Gross Product for All Firms operating in the US | 67 % | 6 % | 25 % |
| Value of Dollar Export Sales by Firms in the BEA Sample (Millions) | 236,371 | NA | NA |
| Coverage of the BeA Sample (in %) Relative to Exports of All Firms operating in the US | 65 % | NA | NA |
| 1994 | | | |
| Total Number of Employees in BEA Sample (Thousands) | 9,622.5 | 2,653.4 | 18,947.4 |
| Gross Product in the BEA Sample (US Millions of Dollars) | 690,466 | 102,520 | 1,325,945 |
| Coverage of the BEA Sample (in %) Relative to Gross Product for All Firms operating in the US | 59 % | 8 % | 26 % |
| Value of Dollar Export Sales by Firms in the BEA Sample (Millions) | 337,036 | NA | NA |
| Coverage of the BeA Sample (in %) Relative to Exports of All Firms operating in the US | 59 % | NA | NA |
| 1999 | | | |
| Total Number of Employees in BEA Sample (Thousands) | 7,954.9 | 2,220,174 | 23,006.8 |
| Value of Dollar Export Sales by Firms in the BEA Sample (Millions) | 441,587 | NA | NA |
| Coverage of the BeA Sample (in %) Relative to Exports of All Firms operating in the US | 62.5 % | NA | NA |

APPENDIX TABLE A.2:
CROSS CHECKING THE ACCURACY OF THE BEA DATABASE

| | Imposing a Cut-off (Reporting Requirement of a Balance Sheet Total of at least 7 Million Euros for Germany, US reporting requirements vary over time, no reporting requirement for Sweden) | Imposing no Cut-off on Germany affiliate reporting |
|--|--|--|
| BEA Data | | |
| Employees of US Affiliates in 1999 in Germany | 458,744 | NA |
| Employees of US Affiliates in 1999 in Sweden | 67,044 | NA |
| German Government Data (Direct US Ownership only) | | |
| Employees of US Affiliates in 1998 | 466,941 | 488,866 |
| Employees of US Affiliates in 1999 | 509,537 | 532,594 |
| Employees of US Affiliates in 2000 | 488,157 | 509,176 |
| Swedish Government Data | | |
| Employees of US Affiliates in 1997 (Majority owned only) | 51,138 | NA |
| Employees of US Affiliates in 1998 (Majority owned only) | 61,089 | NA |
| Employees of US Affiliates in 1999 (Majority owned only) | 78,621 | NA |

APPENDIX TABLE A.3:

DESCRIPTION OF VARIABLES AND DATA SOURCES

| Variable Name | Source | Description |
|--|-----------------------------------|---|
| Log Wage (Industry level) | US Bureau of Economic Analysis | Wages and salaries of employees and employer expenditures for all employee benefit plans in parents computed separately for parents, high-income affiliates and other affiliates and averaged across industries. |
| Log Wage (Industry level) | UNIDO | Wages calculated based on surveys administered by UNIDO, supplemented with secondary sources (such as national statistical agencies). Wages calculated as compensation divided by number of employees at the 3 digit ISIC level Revision 2. All values converted to US dollars using the IMF exchange rate series rf. Data taken from INDSTAT3, published in 2006 by UNIDO. |
| Log Employment | US Bureau of Economic Analysis | Log of the number of full-time and part-time employees on the payroll at the end of the fiscal year in all affiliates. However, a count taken during the year was accepted if it was a reasonable proxy for the end-of-year number. Computed separately for parents, high-income affiliates and other affiliates. |
| R&D Share R&D Share (High-income Affiliates) R&D Share (Low-income Affiliates) | US Bureau of Economic Analysis | Number of employees in research and development as a percentage of total employment. Computed separately for US parents, affiliates in high-income locations and affiliates in low-income locations. |
| US Investment Price | NBER Manufacturing Database | This is the variable PIINV in the NBER's manufacturing productivity database. It is set to 1 in 1987. It combines separate deflators for structures and equipment, based on the distribution of each type of asset in the industry. This is a deflator for new investment flows, not the existing capital stock. See www.nber.org . |
| Foreign Investment Price | Penn World Tables | PPP price of domestic investment calculated from the PWT 6.1. See Appendix for PWT 6.1 for more details, or http://pwt.econ.upenn.edu . |
| Foreign Consumer Goods Price | Penn World Tables | PPP price of consumption goods calculated from the PWT 6.1. See Appendix for PWT 6.1 for more details, or http://pwt.econ.upenn.edu . |
| US Import Penetration | Bernard, Jensen and Schott (2006) | Imports into the US divided by imports into the US plus total production in the US less exports from the US by year by 4-digit SIC 1987 revision code industrial classification. |
| US Import Penetration from Low-Income Countries | Bernard, Jensen and Schott (2006) | Share of products in an industry sourced from at least one country with less than 5 percent of US per capita GDP |

APPENDIX TABLE A.4
 WITHIN ESTIMATES OF LABOR DEMAND BY US PARENTS
 WITH INDUSTRY SALES ADDED AS A CONTROL

| | (1) Pooled | (2) Vertical | (3) Horizontal |
|--|---------------------|---------------------|---------------------|
| Log US Industrial Wages | -0.533 [0.070]** | -0.501 [0.092]** | -0.658 [0.096]** |
| Log Industrial Wages In Low-income Countries | 0.064 [0.033] | -0.315 [0.007]** | 0.273 [0.005]** |
| Log Industrial Wages in High-income Countries | 0.030 [0.023] | -0.001 [0.029] | 0.077 [0.036]* |
| Log of the US Price of Capital | 0.137 [0.161] | 0.319 [0.106]** | 0.262 [0.433] |
| Log of the Foreign Price of Capital | -0.298 [0.149]* | -0.141 [0.204] | -0.487 [0.211]** |
| Log of the Foreign Price of Consumer Goods | 0.734 [0.148]** | 0.536 [0.197]** | 1.034 [0.207]** |
| Import Penetration | -0.098 [0.120] | -0.523 [0.156]** | 0.296 [0.194] |
| Import Penetration from Low Wage Countries | -0.046 [0.205] | 0.524 [0.292] | -0.551 [0.276]** |
| R&D Spending (% Sales) | -0.494 [0.099]** | -0.344 [0.114]** | -1.016 [0.194]** |
| R&D Spending (% Sales) in High-income Countries | 0.308 [0.163] | 0.210 [0.189] | 0.637 [0.310]* |
| R&D Spending (% Sales) in Low-income Countries | 1.021 [0.667] | 1.933 [0.937]* | 0.899 [0.979] |
| Dummy Equal to One if Firm has Employees in Low-income Countries | 0.466 [0.098]** | 0.955 [0.142]** | 0.160 [0.129] |
| Log of Parent Sales by Industry | 0.144 [0.016]** | 0.142 [0.025]** | 0.160 [0.019]** |
| Log of Affiliate Sales by Industry | 0.029 [0.005]** | 0.036 [0.007]** | 0.124 [0.007]** |
| Time dummy 1989 | -0.017 [0.028] | -0.033 [0.034] | 0.019 [0.065] |
| Time dummy 1994 | 0.012 [0.041] | 0.025 [0.047] | 0.039 [0.107] |
| Time dummy 1999 | 0.101 [0.046]* | 0.102 [0.052]* | 0.154 [0.120] |
| Observations | 3866 | 3866 | 3866 |
| Number of firms | 1868 | 1868 | 1868 |
| R-squared | 0.07 | 0.18 | 0.18 |

Standard errors in brackets

* significant at 5%

** significant at 1%

APPENDIX TABLE A.5

TESTING FOR THE IMPACT OF SELECTION INTO EXIT: HECKMAN CORRECTION

| | (1) | (2) | (3) |
|--|-----------------------------|---------------------|---------------------|
| | Pooled (All Enterprises) | Vertical | Horizontal |
| Log US Industrial Wages | -0.554 [0.100]** | -0.422 [0.113]** | -0.890 [0.128]** |
| Log Industrial Wages in Low-income Countries | 0.051 [0.065] | -0.228 [0.110]* | 0.136 [0.074]* |
| Log Industrial Wages in High-income Countries | -0.013 [0.026] | 0.005 [0.030] | 0.113 [0.039]* |
| Import Penetration | -0.395 [0.155]* | -0.146 [0.007]** | -0.020 [0.255] |
| Import Penetration from Low Wage Countries | 0.375 [0.299] | 1.091 [0.388]** | -0.380 [0.098]* |
| R&D (% Employment) | -0.204 [0.107] | 0.087 [0.113] | -2.100 [0.272]** |
| R&D (% Employment) in High-Income Countries | 1.575 [0.410]** | 1.569 [0.431]** | -0.069 [0.966] |
| R&D Spending (% Employment) in Low-income Countries | -1.419 [1.127] | 0.893 [0.275]** | -0.030 [1.562] |
| Dummy Equal to One if Firm has Employees in Low-income Countries | -0.013 [0.147] | 0.445 [0.232]* | -0.277 [0.080]* |
| Lamda for 1994 | -0.037 [0.091] | 0.051 [0.102] | -0.243 [0.122]* |
| Lamda for 1999 | 0.238 [0.068]** | 0.392 [0.078]** | -0.204 [0.095]** |
| Number of Observations | 2319 | 2319 | 2319 |
| R-Square | 0.05 | 0.12 | 0.12 |

Notes: All specifications Include Time Dummies and Time Dummies interacted with year-specific inverse Mills Ratio Terms. Other controls not reported include consumption prices and capital goods prices. Excluded determinants of selection, not included in second stage, include lagged profits. Selection equation also includes all independent variables in second stage. Standard errors in parentheses.