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Global Production Sharing and Rising Inequality: A Survey of Trade and Wages*

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Running Head – Global Production

We argue that trade in intermediate inputs, or “global production sharing,” is a potentially important explanation for the increase in the wage gap between skilled and unskilled workers in the US and elsewhere. Using a simple model of heterogeneous activities within an industry, we show that trade in inputs has much the same impact on labor demand as does skill-biased technical change: both of these will shift demand away from low-skilled activities, while raising relative demand and wages of the higher skilled. Thus, distinguishing whether the change in wages is due to international trade, or technological change, is fundamentally an empirical rather than a theoretical question. We review three empirical methods that have been used to estimate the effects of trade in intermediate inputs and technological change on wages, and summarize the evidence for the US and other countries.

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1. Introduction

One of the most widely-discussed public policy issues in the United States and many other industrial countries is the decline in the wages of less-skilled workers during the 1980's and 1990's, both in real terms and relative to the wages of more-skilled workers. The question is, what factors account for this change? One obvious explanation that comes to mind is increased competition from low-wage countries. Surprisingly, many economists researching this issue have come to the conclusion that trade is not the dominant – or even an important – explanation for the shift in wages. They have instead looked to the massive influx of computers into the workplace, and other forms of technological change, as the explanation.

In this survey, we present a contrary point of view, and argue that international trade is indeed an important explanation for the increase in the wage gap. Our argument rests on the idea that an increasing amount of international trade takes the form of trade in intermediate inputs. This is sometimes called “production sharing” by the companies involved, or simply “outsourcing.”¹ Trade of this type affects labor demand in import-competing industries, but also affects labor demand in the industries using the inputs. For this reason, trade in intermediate inputs can have an impact on wages and employment that is much greater than for trade in final consumer goods. As we shall argue, trade in inputs has much the same impact on labor demand as does skill-biased technical change: *both* of these will shift demand away from low-skilled activities, while raising relative demand and wages of the higher skilled. Thus, distinguishing whether the change in wages is due to international trade, or technological change, is fundamentally an empirical rather than a theoretical question.

In the next section, we review the basic evidence that has been used to conclude that trade has not been a significant cause of US wage changes. We argue that this evidence still leaves

room for trade to be important, especially trade in intermediate inputs. Empirically, a good deal of trade is in intermediate inputs, and the impact of this on wages and other factor prices is quite different from that obtained with just trade in final goods. This is shown in section 3, where we present a simple model of trade in intermediate inputs. In section 4 we discuss various methods of estimating this model, and summarize the evidence for the US and other countries.

Conclusions and directions for further research are given in section 5.

2. Changes in Wages and Employment

The basic facts concerning wage movements in the United States are fairly well understood.² For full-time US workers between 1979 and 1995, the real wages of those with 12 years of education fell by 13.4% and the real wages of those with less than 12 years of education fell by 20.2%. During the same period, the real wages of workers with 16 or more years of education rose by 3.4%, so that the *wage gap* between less-skilled and more-skilled workers increased dramatically.³ To illustrate these trends, we can use data from the US manufacturing sector for “nonproduction” and “production” workers. The former are often used as a proxy for more-skilled workers, and the latter as a proxy for less-skilled workers.⁴ These trends are shown in Figure 1, which graphs the relative annual earnings of nonproduction/production workers in US manufacturing, and Figure 2, which graphs their relative annual employment.

In Figure 1, we see that earnings of nonproduction relative to production workers in the US moved erratically during the 1960’s and 1970’s, but then increased substantially during the 1980’s and 1990’s. Turning to Figure 2, we see that there has been a steady increase in the ratio of nonproduction to production workers used in US manufacturing, with some leveling off recently. This increase in the supply of workers can account for the *reduction* in the relative

wage of nonproduction workers from 1970 to the early 1980s, as shown in Figure 1, but is at odds with the *increase* in the relative nonproduction wage after that (Katz and Murphy, 1992). The rising relative wage should have led to a shift in employment *away* from skilled workers, along a demand curve, but it has not. Thus, the only explanation consistent with the facts is that there has been an *outward shift* in the demand for more-skilled workers since the mid-1980s, leading to an increase in their relative employment and wages (Katz and Autor, 1999).

The same decline in the relative wages of blue-collar workers during the 1980's and into the 1990's can be found for Australia, Canada, Japan, Sweden, and the United Kingdom (Freeman and Katz, 1994; Katz and Autor, 1999),⁵ and also for Hong Kong and Mexico (Cragg and Epelbaum, 1996; Hanson and Harrison, 1999; Hsieh and Woo, 1999). What factors account for these changes? Most widely cited are international competition from low-wage countries and skilled-biased technological change due to the increased use of computers, with the latter considered as the most important.⁶ There are at least three reasons why trade is thought to have played a rather minor role, and these are reviewed in the next sections.

The Magnitude of Trade

First, it is often noted that the magnitude of trade flows to and from the United States, especially with developing countries, is too small to lead to the observed wage changes. Indeed, for many industrial countries, the ratio of trade to gross domestic product (GDP) in 1970 was no higher than it was just before World War I.⁷ In the U.S., for example, the value of trade (an average of imports and exports) was 6.1% of GDP in 1913, but only 4.1% in 1970, rising to 8.8% in 1980. Other industrial countries have higher levels of trade, but many (including France, Germany, Italy, and Sweden) show the same time-pattern as in the US. A few other countries,

such as Australia, Denmark, Japan and the United Kingdom, still have not reached the trade/GDP ratio that they had in 1913. Krugman (1995, p. 331) uses these observations to conclude that: “it would be hard to argue that the sheer volume of trade is now at a level that marks a qualitative difference from previous experience.”⁸

But the ratio of trade to GDP does not tell the whole story. All industrial countries have had increasing shares of their economies devoted to services rather than merchandise (i.e., manufacturing, mining and agriculture). To make a better comparison of trade with overall production, we should measure merchandise goods in both the numerator and the denominator – i.e., compare merchandise trade to merchandise value-added. When this is done, there are still two countries for which this ratio was larger in 1913 than in 1990 (Japan and the United Kingdom) and one other for which this ratio changed little (Australia). But all other industrial countries have experienced *substantial growth* in trade relative to merchandise value-added between 1913 and 1990: this ratio has increased by about one-third for Denmark and Norway; by three-quarters for Canada; has doubled for France, Germany, Italy, and Sweden; and has nearly tripled for the United States, rising from 13.2% in 1913 to 35.8% in 1990. We conclude that merchandise trade has indeed grown substantially relative to the production of these commodities in many advanced countries.

Has the composition of merchandise trade changed over time? Various evidence indicates that intermediate goods play an increasingly important role in trade. One approach is to look at “processing trade,” which is defined by customs offices as the import of intermediate inputs for processing, and subsequent re-export of the final product. This activity has grown enormously in China, for example, for which Hong Kong often serves as an intermediary. For

example, between 1988 and 1998, processing exports grew from \$12.4 billion to \$97.2 billion, or from about one-third to over one-half of total Chinese exports (Feenstra and Hanson, 2001, Table 1). This outward processing serves newly industrialized countries in Asia, but also developed countries such as the U.S., Japan and Europe. Between the industrialized countries, too, there has been an increase in processing trade. Görg (2000, Table 1) reports on the increase in US processing trade with the European Union between 1988 and 1994. He finds that US processing imports into these countries (as a share of their total US imports) increased slightly from 17.7% to 19.8%, but this same ratio increased more significantly from 13.7% to 23.7% for the “periphery” countries of Greece, Ireland, Portugal and Spain.

In addition to processing trade, the *total* amount of imported intermediate inputs can be estimated by using the purchases of each type of input, and multiplying this by the economy-wide import share for that input. Summing this over all inputs used within each industry, we obtain estimated imported inputs, which can then be expressed relative to total intermediate input purchases. Feenstra and Hanson (1999) perform this calculation for US manufacturing industries, and find that imported inputs have increased from 6.5% of total intermediate purchases in 1972 to 8.5 % in 1979, and 11.6% in 1990.

Campa and Goldberg (1997) make the same calculation for Canada, Japan, the United Kingdom, and the United States, and their results are shown in Table 1.⁹ The United States shows a doubling of the share of imported inputs between 1975 and 1995 for all manufacturing, though it is still at a low level compared to Canada and the United Kingdom, where over 20% of inputs were purchased from abroad in 1993. The United Kingdom, especially, shows a large absolute increase in foreign outsourcing. For individual industries, the chemical industry has a lower share of imported inputs than overall, whereas machinery (non-electric and electric) and

transportation equipment have higher shares in these three countries. The machinery and transportation industries have especially rapid growth in imported inputs, with the shares doubling or even tripling between 1974 and 1993. The exception to these observations is Japan, where the share of imports in these heavy industries is lower than in overall manufacturing, and has generally been falling. With this single exception, the increased use of imported inputs is a characteristic feature of many industrial countries over the past two decades.

Changes in Import Prices

The second reason why some authors have argued that international trade is not a significant factor in explaining the movement in wages has to do with the behavior of import and export prices. In widely cited work, Lawrence and Slaughter (1993) have shown that the movement of prices across industries seems to contradict the movement of relative wages.¹⁰ In order for international competition to be the cause of the fall in the relative wage of less-skilled workers, we should see that prices of the least-skill-intensive goods – such as apparel – have fallen relative to other goods. While relative prices for apparel goods fell in the 1970's, they were stable in the 1980's (Leamer, 1998). Prices for other less-skilled-intensive goods actually rose in the 1980's.

This can be seen from Table 2, which is taken from the work of Lawrence and Slaughter (1993) and Lawrence (1994). For each country, the first row is a weighted average of the change in manufacturing prices over the 1980s, where the weights are the industry's share of total manufacturing employment of *nonproduction* workers. The second row is again the weighted average of the change in prices over the 1980s, but now using the industry's share of employment of *production* workers. For US import prices, for example, we can see that when

industries are weighted by their production workers, the average price increase is *higher* than when we weight by non-production workers. The same pattern can be seen by comparing the rows for other industrial countries. This means that some of the industries that use the *most* production – or less-skilled – workers are those with the *highest* price increases. This finding led Lawrence and Slaughter (1993) to conclude that the price movements, due to international competition, could not explain the wage movements.

However, if we accept that industries are engaged in importing intermediate inputs, then this suggests a different way to look at the price data. Rather than comparing prices *across* different industries, depending on their skill-intensity, it now makes sense to compare import and domestic prices *within* each industry. The types of goods being imported within each industry (e.g., auto parts) are not the same as those being sold domestically (e.g., finished autos). Indeed, as US firms find imported inputs at increasingly lower prices – through outsourcing activities that they used to do at home – we would expect to see that US prices within each industry should be *rising* relative to import prices. In terms of Table 2, we should be comparing the price changes across columns rather than across rows. We see that for the United States during the 1980s it is indeed the case that domestic prices rose faster than import prices, and the same is true for Japan and Germany. These price movements are entirely consistent with a model of foreign outsourcing, whereby the United States and other industrial countries are continually seeking lower-cost sources of supply. Based on this logic, there is no “contradiction” at all between the movement of prices and relative wages.

Employment Changes Within and Between Industries

The third piece of empirical evidence comes from decomposing the shifts in the relative employment of less-skilled workers into those occurring *within* industries, and those occurring *between* industries. According to this line of reasoning, international trade should have the effect of moving workers *between* sectors, as industries expand or contract in response to foreign competition. In contrast, new technology, such as the increased use of computers, would change the ratio of more-skilled to less-skilled workers employed *within* each sector. Some evidence on this *within versus between* industry distinction is contained in Table 3, which is taken from Berman, Bound and Griliches (1994), and Bernard and Jensen (1997).

Part A of Table 3 decomposes the change in the relative employment and relative wages of nonproduction workers into those that occurred within and between industries. We can see that in the period 1979-1987, the relative employment of nonproduction workers increased by slightly more than one-half of one percent per year (0.546%), with about two-thirds of that (0.362%) explained by *within* industry movements. On the wages side, the relative annual earnings of nonproduction workers increased by about seven-tenths of a percentage point per year (0.719%), with more than half of that change (0.410%) explained by *within* industry movements. The conclusion suggested by Berman, Bound and Griliches (1994) is that trade cannot be a dominant explanation for the wage and employment shifts, because the *between* industries movements are smaller than the *within* industry movements.

However, that conclusion raises the question of what is occurring *within* these industries, and whether that shift could itself be related to international trade. Bernard and Jensen (1997) have obtained some suggestive evidence on this point, by doing the same decomposition but using *plant-level* data rather than *industry-level* data. This is shown in Part B of Table 3. Looking again at the period 1979-1987, we can see that nearly one-half of the relative increase in

the employment of non-production workers (0.392%) occurred due to shifts *between* plants (0.177%) , and more than one-half of the increase in the relative wage of non-production workers (0.536%) is also explained by movements *between* manufacturing plants (0.315%). Furthermore, Bernard and Jensen have found that the plants experiencing the greatest increase in relative nonproduction employment and earnings are precisely those that are engaged in export activity.

The results of Bernard and Jensen provide *prima facie* evidence that trade has had an impact on factor demand and wages. In order to understand what these linkages are, we present in the next section a simple model of outsourcing, that builds upon the key insight of their empirical work: the heterogeneity of production activities within industries.

3. A Simple Model of Outsourcing

Of the many activities that take place within any industry, let us identify just three: the production of an unskilled-labor intensive input, denoted by y_1 ; the production of a skilled-labor intensive input, denoted by y_2 , and the “bundling together” of these two goods into a finished product. The two intermediate inputs are produced at home and also traded internationally; by including such trade in intermediate inputs, our model therefore incorporates outsourcing or “production sharing.” We shall simplify the analysis, however, by assuming that the production of these two inputs and the “bundling” activity are *always* performed at home; thus, we are ruling out corner solutions where one of these activities is done entirely abroad. In reality, corner solutions such as this are very common. For example, many US firms export intermediate inputs to the *maquiladora* plants in Mexico, where assembly of the inputs and other production activities take place there rather than in the U.S.¹¹ A model of production sharing that emphasizes the movement of entire activities across countries is developed by Feenstra and

Hanson (1996). Our theoretical treatment in this survey is simplified by only examining the *marginal* movement of production across countries, as induced by changes in prices. Despite this simplification, we will be able to motivate much of the empirical work in this area.

We note that theoretical models of production sharing are only starting to take hold within international economics: in addition to Feenstra and Hanson (1996), examples include the papers gathered in Arndt and Keirzkowski (2001), as well as Kim and Mieszkowski (1995), Leamer (1996), Xu (2000), Yi (2000), and Marjit and Acharyya (forthcoming).¹² This concept is also used in economic sociology (Gereffi and Korzeniewicz, 1994), geography (Dicken *et al*, 2001; Yeung, 2001) and other social sciences, where production sharing is referred to by the more general name “commodity chains.” A commodity chain consists of the sequence of activities involved in the manufacture of a product, from initial development through to production, marketing and sales, especially as these activities cross international boundaries. In these disciplines, commodity chains are seen as an integral part of the development process for countries that are still industrializing, i.e. a country’s position in the commodity chain will impact its standard of living. While we will be taking a less grand view, it will still be the case in our simplified model that production sharing has a substantial impact on wages.

We will suppose that the two inputs y_i , $i=1,2$, are each produced using unskilled labor (L_i), skilled labor (H_i), and capital (K_i), with concave and linearly homogeneous production functions,

$$y_i = f_i(L_i, H_i, K_i), \quad i=1,2. \quad (1)$$

For example, the unskilled-labor intensive input y_1 might represent the activities done within a factory, while the skilled-labor intensive input y_2 might represent the research and development

activities within the industry, as well as marketing and after-sales service. These are both needed to produce the final manufacturing product. But some of the activities done within the factory can instead be outsourced, i.e. imported from abroad; and conversely, the services associated with research, development and marketing can be exported to support production activities abroad. We therefore let $x_1 < 0$ denote the imports of input 1, and $x_2 > 0$ denote the exports of input 2. For convenience, we choose the exported intermediate input as the numeraire, so we will hold this price fixed (at unity), and let p denote the price of the imported input x_1 .

The production of the final manufacturing good is given by $y_m = f_m(y_1 - x_1, y_2 - x_2)$, where this production function “bundles together” the amounts of goods 1 and 2 available, and is also concave and linearly homogeneous. We ignore any additional labor and capital inputs used in this bundling activity, so that the total factor usage in the manufacturing industry is,

$$L_1 + L_2 = L_m, \quad H_1 + H_2 = H_m, \quad K_1 + K_2 = K_m. \quad (2)$$

We can now solve for the optimal output in the industry, which includes the three activities. With perfect competition, the value of output from the final good, plus net trade, will be maximized subject to the resource constraints:

$$F_m(L_m, H_m, K_m, p_m, p) \equiv \max_{x_1, L_1, H_1, K_1} p_m f_m(y_1 - x_1, y_2 - x_2) + p x_1 + x_2, \quad (3)$$

subject to (1), (2),

where p_m is the price of the final good, and p is the price of the imported intermediate input.

Note that the optimization problem in (3) does *not* require that trade is balanced on an industry-by-industry basis, i.e. we do not require that $p x_1 + x_2 = 0$. Of course, there will be some balance

of trade constraint for the economy overall, but we ignore that here. The value of the industry production function in (3) can be thought of as value-added for the industry, i.e. nominal output including exports x_2 less the value of intermediate inputs px_1 .

Problem (3) can be easily illustrated, as in Figure 3, where we show the production possibility frontier between inputs 1 and 2, and several isoquants of the final good y_m . For the purpose of illustration we now add the extra condition that trade in the inputs is balanced ($px_1 + x_2 = 0$), so that the output of the final good is maximized on the isoquant that is tangent to the balanced trade line. At initial prices, for example, the industry produces inputs at A, and then trades to B. With a *drop* in the relative price of the imported input, the industry shifts production towards the skilled-labor intensive activity at A', and then trades to B', obtaining a higher output y_m . All this will look very familiar to most readers: the only special feature of Figure 3 is that we think of these activities taking place *within* a single manufacturing industry. Let us define Y_m as the optimized value of (3), which measures measured value-added in industry m ,

$$Y_m \equiv F_m(L_m, H_m, K_m, p_m, p) \quad . \quad (4)$$

Provided that the underlying production functions f_i , $i=1,2,m$, are increasing and concave, then the function F_m will also be increasing and concave in (L_m, H_m, K_m) . We can think of F_m as an “aggregate production function” for the industry.

The difficulty with using this “aggregate production function” for any empirical or theoretical work is that it implicitly holds fixed the level of labor and capital used in the industry, i.e. at those levels given by the constraints in (2). Instead, we would like to think of labor, and possibly capital too, as being optimally adjusted in response to changes in factor prices. To

reflect this, we make use of the *cost function* that is dual to (4). First, we can define a *short-run cost function*, obtained when the level of capital and output are fixed:

$$C_m(w, q, K_m, Y_m, p, p_m) \equiv \min_{L_m, H_m} wL_m + qH_m, \text{ subject to (4),} \quad (5)$$

where w is the wage of unskilled labor and q is the wage of skilled labor. Alternatively, we can define a *long-run cost function*, obtained when labor and capital are both chosen optimally:

$$C_m(w, q, r, Y_m, p, p_m) \equiv \min_{L_m, H_m, K_m} wL_m + qH_m + rK_m, \text{ subject to (4).} \quad (6)$$

where r is the rental on capital.

Both of these cost functions have been used empirically, as we discuss in the following sections. Before turning to this material, however, it is well worth exploring a few theoretical properties. In particular, if we add the condition that marginal cost equals product prices, and then allow prices to change, what will be the impact on factor prices? In Figure 3, for example, we showed how a fall in the relative price of imports p would raise final output y_m , but what is the corresponding impact on factor prices? In the remainder of this section we explore this question theoretically, making use of the long-run version of the cost function.

If both types of labor and capital are being optimally chosen for the industry overall, as in (6), then they must also be optimally chosen within each of the activities $i=1,2$. So consistent with (6), we can also define the long-run cost functions for the disaggregate activities within the industry:

$$C_i(w, q, r, Y_i) \equiv \min_{L_i, H_i, K_i} wL_i + qH_i + rK_i, \text{ subject to (1), for } i=1,2. \quad (7)$$

Since the activity production functions in (1) are assumed to be linearly homogeneous, then the activity-level cost functions in (7) will be homogeneous of degree one in Y_i . This implies they can be written as $C_i(w, q, r, Y_i) = Y_i c_i(w, q, r)$, where $c_i(w, q, r)$ is the *unit-cost* function (equal to marginal cost or average cost).

Then the zero-profit conditions for activities 1 and 2 can be written as:

$$p = c_1(w, q, r), \quad 1 = c_2(w, q, r), \quad (8)$$

These conditions must hold in order for the locally produced inputs y_i , $i=1,2$, to be competitive with those available from abroad, at the prices p and unity, respectively. Totally differentiating (8) using the familiar Jones' (1963) algebra, we can express the percentage change in factor prices \hat{w} , \hat{q} , and \hat{r} as functions of the percentage change in the import price \hat{p} :

$$\hat{p} = \theta_{1L} \hat{w} + \theta_{1H} \hat{q} + \theta_{1K} \hat{r}, \quad 0 = \theta_{2L} \hat{w} + \theta_{2H} \hat{q} + \theta_{2K} \hat{r}, \quad (9)$$

where θ_{ij} is the cost-share of factor j in activity i , with $\sum_j \theta_{ij} = 1$. Treating the change in the import price \hat{p} as exogenous, (9) gives *two* equations with which to determine *three* unknown factor prices changes – \hat{w} , \hat{q} , and \hat{r} . In general, these factor price changes will be difficult to pin down with only two equations. In terms of Figure 3, when production shifts towards the skilled labor intensive activity, from A to A' , we do not know in general how factor prices are affected. But there are some simplifying assumptions we can make which allow us to determine these.

Case 1 – Equal Cost Shares of Capital

As in Feenstra and Hanson (1996), we can assume that capital has equal cost shares in the two industries, so that $\theta_{1K} = \theta_{2K}$. Using this, we take the difference between the two

equations in (9) to obtain,

$$\hat{p} = (\theta_{1L} - \theta_{2L})\hat{w} + (\theta_{1H} - \theta_{2H})\hat{q} = (\theta_{1L} - \theta_{2L})(\hat{w} - \hat{q}), \quad (10)$$

where the second equality follows since with equal cost shares of capital, the total cost shares of labor are also equal, so that $(\theta_{1L} + \theta_{1H}) = (\theta_{2L} + \theta_{2H}) \Rightarrow (\theta_{1L} - \theta_{2L}) = -(\theta_{1H} - \theta_{2H})$. With activity 1 assumed to be unskilled-labor intensive, we have that $(\theta_{1L} - \theta_{2L}) > 0$. Thus, (10) says that a *decrease* in the price of imported intermediate input, $\hat{p} < 0$, leads to a *decrease* in the relative wage of unskilled labor, $(\hat{w} - \hat{q}) = \hat{p}/(\theta_{1L} - \theta_{2L}) < 0$.

These results are illustrated in Figure 4, where we graph the iso-curves of $p = c_1(w, q, r)$ and $1 = c_2(w, q, r)$. With the wages of unskilled and skilled labor labeled on the axis, we are implicitly holding fixed the rental on capital, r . Now suppose that there is a fall in the price p of activity 1, which is unskilled labor intensive. In Figure 4, this will shift inwards the iso-cost line of that activity, and as shown, will lead to a *fall* in the relative wage of unskilled labor (from A to B). There will be some additional change in the rental on capital, but under our assumption of equal cost share of capital in the two industries, this will lead to an *equi-proportional* shift in the two iso-cost curves and therefore have no further effect on the relative wage. Thus, the drop in the price of the imported inputs leads to a fall in the relative wage (w/q) of unskilled labor.

We can also ask what happens to the price of the final good p_m . Let $c_m(p, 1)$ denote the unit-cost function that is dual to $f_m(y_1, y_2)$, where recall that the price of activity 2 is unity. Then the price of the final good satisfies $p_m = c_m(p, 1)$, so that $\hat{p}_m = \theta_{m1}\hat{p}$, where θ_{m1} is the cost-share of input 1 in the final product. Thus, with a fall in the price of imported inputs, $\hat{p} < 0$, the price of the final good also falls but by *less*, $\hat{p} < \hat{p}_m < 0$. Stated differently, the price of the final good

relative to imported inputs *rises*, $\hat{p}_m - \hat{p} > 0$. This in fact is what happened in the US and other industrial countries, as shown in Table 2, where the change in domestic prices exceeds the change in import prices over 1980-1990. Our theoretical model therefore confirms that this price movement is consistent with the fall in the relative wage of unskilled labor.

Case 2 – *The Unskilled-Labor Intensive Activity is also Capital Intensive*

A second case, emphasized by Sachs and Shatz (1998), is where activity 1 uses more unskilled labor, $(\theta_{1L} - \theta_{2L}) > 0$, and also more capital, $(\theta_{1K} - \theta_{2K}) > 0$; think of factory production, for example. They suppose that the price of the imported input is *constant*, $\hat{p} = 0$, but that the rental price on capital *increases*, $\hat{r} > 0$, because capital is leaving the country. In Figure 5, this will lead to an inward shift in both iso-cost curves, but since activity 1 uses more capital, a *larger* shift in that curve. Thus, the factor-price equilibrium will shift from A to B in Figure 5, where the wage of unskilled labor has fallen. We are not sure whether the wage of skilled labor rises or falls, but in any case, we obtain a fall in the relative wage (w/q) of unskilled labor.¹³ Intuitively, because activity 1 uses both unskilled labor and capital intensively, when capital leaves it is then unskilled labor that suffers. In this case, the cause of the change in wages is an outflow of capital from the country.¹⁴

Notice that the change in relative wages occurs without any change in the price of the imported input or the final good. We might expect, though, that there will be some impact on the price of a nontraded good. Denoting the zero-profit condition for a nontraded good by $p_n = c_n(w,q)$, we can totally differentiate this expression to obtain:

$$\hat{p}_n = \theta_{nL} \hat{w} + \theta_{nH} \hat{q}, \quad (11)$$

where θ_{nj} is the cost-share of factor $j=L,K$. For simplicity, we assume the nontraded good uses no capital in production. In terms of Figure 5, the nontraded good would have an additional iso-cost line (not drawn) that goes through both points A and B. The question we wish to determine is whether the change in factor prices from A to B implies an *increase* or a *decrease* in the price of the nontraded good.

To answer this, let us consider two extreme cases: where the non-traded good uses only unskilled labor, or only skilled labor. In the first case, the iso-cost line for the nontraded good would be vertical in Figure 5, i.e. each level of the unskilled wage w would correspond to a unique value for costs $c_n(w)$. In that case, the movement of the factor-price equilibrium from A to B would require a shift in the iso-cost line of the non-traded good to the left, and a corresponding *fall* in the price of the nontraded good (exactly like the fall in unskilled wages). Alternatively, if the nontraded good uses only skilled labor, then its iso-cost line would be horizontal in Figure 5, so that each level of the skilled wage q corresponds to a unique value for costs $c_n(q)$. In this case, we are *unsure* whether the movement of the equilibrium from A to B implies a rise or fall in the price of the nontraded good. If the iso-cost line of activity 1 shifts down sufficiently more (i.e. activity 1 is much more capital intensive than activity 2), then it is visually apparent that point B would imply a *higher* wage for skilled labor, and a *higher* price for the nontraded good if it uses enough of this factor. In fact, Sachs and Shatz (1998) argue for the US that nontraded goods are indeed skilled-labor intensive relative to manufacturing overall. In that case, the exit of capital, and increase in the relative wage of skilled labor, can be expected to lead to a *rise* in the price of the nontraded good. Conversely, if the nontraded good is more unskilled-labor intensive, then its price will *fall*.

The evidence for the US is that skilled-labor intensive nontradables experienced a *rise* in price during the 1980s, whereas unskilled-labor intensive nontradables experienced a *fall* in prices, consistent with these theoretical predictions. As we shall discuss at the end of the next section, Harrigan and Balaban (1999) and Harrigan (2000) find that these changes in nontradables prices are highly correlated with change in relative wages, but this leaves open the question of *causality*: are the prices of nontradables driving the relative wages or, as we have suggested here, are the relative wages driving the prices of nontradables? We have shown that the exit of capital in an open economy can quite plausibly have the effect of raising *both* the relative wage of skilled labor and the price of skill-intensive nontradables, and lowering *both* the relative wage of unskilled labor and the price of unskilled-intensive nontradables, which is consistent with evidence for the US during the 1980s.

4. Results from Empirical Studies

Summarizing our argument so far, the decision of companies to purchase intermediate inputs from overseas will most certainly affect their employment at home, and can be expected to differentially affect skilled versus unskilled workers. With firms in industrial countries facing a higher relative wage for unskilled labor than that found abroad, the activities that are outsourced would be those that use a large amount of unskilled labor, such as assembly of components and other repetitive tasks. Moving these activities overseas will reduce the relative demand for unskilled labor in the industrial country, in much the same way as replacing these workers with automated production. *This means that outsourcing has a qualitatively similar effect on reducing the relative demand for unskilled labor within an industry as does skilled-biased technological change, such as the increased use of computers.* Thus, determining which of

these is most important is an empirical question. We will examine three methods that have been used to estimate the effects of trade versus technological change on wages and employment.

Estimation of Demand for Skilled Labor

The first empirical method estimates industry production functions, such as (4), and attempts to determine which factors affect the relative demand for skilled labor. The starting point is to consider the short-run cost function that is dual to the industry production function, $C_m(w, q, K_m, Y_m, p, p_m)$, as derived in (5). Notice that we have included the price of imported inputs p in this cost function, as well as in the industry price p_m , since they also appear in the production function (4). More generally, *any structural variables that shift the production function and therefore affect costs* should be included as arguments. We will denote these variables by the vector z ; in addition to the price of imported inputs, this can include expenditures on computers and other new types of capital equipment. We therefore rewrite the cost function in (5) as $C_m(w, q, K_m, Y_m, z)$.

The next step is to choose a functional form for costs. A convenient choice is the translog cost function, which was introduced by Diewert (1974, p. 139) and initially used in the international trade literature by Kohli (1978, 1991). This function is written in a general notation as,

$$\begin{aligned} \ln C = & \alpha_0 + \sum_{i=1}^I \alpha_i \ln w_i + \sum_{k=1}^K \beta_k \ln x_k + \frac{1}{2} \sum_{i=1}^I \sum_{j=1}^I \gamma_{ij} \ln w_i \ln w_j \\ & + \frac{1}{2} \sum_{k=1}^K \sum_{\ell=1}^K \delta_{k\ell} \ln x_k \ln x_\ell + \sum_{i=1}^I \sum_{k=1}^K \phi_{ik} \ln w_i \ln x_k \end{aligned} \quad (12)$$

where w_i denotes the wages of the optimally chosen inputs $i=1,\dots,I$, and x_k denotes either the quantities of the *fixed* inputs or outputs $k=1,\dots,K$, or *any other shift parameters*. In terms of the cost function in (11), there are just two optimally chosen factors – skilled and unskilled labor – while capital and output are treated as fixed in the short run. In order to ensure that the cost function in (12) is linearly homogeneous in wages, we impose the requirements,¹⁵

$$\sum_{i=1}^I \alpha_i = 1 \quad \text{and} \quad \sum_{i=1}^I \gamma_{ij} = \sum_{j=1}^I \gamma_{ij} = 0. \quad (13)$$

The usefulness of the translog function comes from computing its first derivatives, $\partial \ln C / \partial \ln w_i = (\partial C / \partial w_i)(w_i / C)$. Because $\partial C / \partial w_i$ equals the demand for the chosen input i , it follows that $(\partial C / \partial w_i)(w_i / C)$ equals the *payments to factor i relative to total costs*, which we denote by the cost-shares s_i . Thus, differentiating (12) with respect to $\ln w_i$, we obtain,

$$s_i = \alpha_i + \sum_{j=1}^I \gamma_{ij} \ln w_j + \sum_{k=1}^K \phi_{ik} \ln x_k, \quad i=1,\dots,I. \quad (14)$$

Given annual data on factor cost shares, wages, and fixed inputs and outputs, this set of linear equations can be estimated over time for a given industry to obtain the coefficients γ_{ij} and ϕ_{ik} .¹⁶ Alternatively, the equations can be estimated for a single year, or the change between two years, by pooling data across industries. In the latter case, we are assuming that the *same* cost function applies across the industries. Despite this strong assumption, the cross-industry approach is popular and we shall follow it here.

Returning now to the notation $C_m(w, q, K_m, Y_m, z)$, we have two chosen inputs – skilled and unskilled labor. Focusing on the share equation for skilled labor, it will depend on wages for

both types of labor, as well capital, output, and all other structural variables, z . When (14) is estimated by pooling data across industries, as in Berman, Bound and Griliches (1994) for example, it is felt the cross-industry variation in wages has little information: wages differ across industries principally due to quality-variation of workers, so we do not expect high-wage industries to economize on those (high-quality) workers. Accordingly, the wage terms are typically dropped from the right of (14) when pooling data across industries. This leaves just fixed capital, output, and other structural variables. Taking the difference between two years, the estimation equation for the wage-share of skilled labor (s_{Hm}) in industries $m=1, \dots, M$ becomes:

$$\Delta s_{Hm} = \phi_0 + \phi_K \Delta \ln K_m + \phi_Y \Delta \ln Y_m + \phi_Z' \Delta z_m, \quad m=1, \dots, M, \quad (15)$$

where z_m denotes the vector of structural variables that shift costs, and ϕ_m is the corresponding vector of coefficients. In particular, when the wage-share of skilled labor is increasing, we are interested in determining how much of that increase is due to changes in capital, output, and the structural variables.

Estimates of (15) for 447 industries within the US manufacturing sector, over 1979-1990, are shown in Table 4. The data are from the NBER Productivity Database (Bartelsman and Gray, 1996, which is available at <http://www.nber.org/nberces/>). In these regressions we use nonproduction labor as a proxy for skilled labor, so the dependent variable is the change in the share of nonproduction labor in total wages within each industry. Over all manufacturing industries, the nonproduction wage share increased from 35.4% to 42.4% between 1979 and 1990, for an annual growth of 0.4% per year.

The specifications in Table 4 are similar to those in Berman, Bound and Griliches (1994), Autor, Katz and Krueger (1998), and Feenstra and Hanson (1999). They include as regressors:

the shipments of each industry (as a proxy for value-added Y_m); the capital/shipments ratio (reflecting the capital input K_m relative to Y_m); foreign outsourcing, measured by imported intermediate inputs as a share of total materials purchases (this term is used instead of the *price* of imported inputs p , since the prices are not observable); the shares of computers and other high-tech capital in the total *capital stock*, and the share of expenditures on computers in *total investment* (both these variables can be viewed as components of the overall capital stock K_m). The share of computers and other high-tech capital in the total *capital stock* is constructed using either *ex post* rental prices, or *ex ante* rental prices.¹⁷ In contrast, the share of computer spending in *investment* is obtained from *the Census of Manufactures*, which simply asked firms to report what percentage of new investment was devoted to computers. This variable has been used previously by Berman, Bound and Griliches (1994) and Autor, Katz and Krueger (1998). We feel that measuring computers as a share of the total *capital stock* is preferable. All variables are at the industry level and all, except the computer investment share, are measured as annual changes. All regressions are weighted by the industry share of the total manufacturing wage bill.

In column (1) of Table 4, we report the mean values of the dependent and independent variables for 1979-1990, and, following this, the regression coefficients in columns (2)-(4). Each regression uses alternative measures of the computer and other high-technology shares. In all the regressions, we see that outsourcing has a positive impact on the nonproduction share of the wage bill, as does the computer share. By multiplying the regression coefficients by the mean values for the change in each variable, we obtain the contributions of each to the total change in the nonproduction wage share, shown in column (5). We see that outsourcing accounts for 15-24% of the shift towards nonproduction labor.¹⁸

The results for computers depend on the specification. Measuring computer services and other high-tech capital as a share of the capital stock, using *ex post* rental prices, we see they account for 13% of the shift towards nonproduction labor. Measuring these shares using *ex ante* rental prices, we see that that computers and other high-tech capital explain only 8% of this shift. In both cases, the contribution of computers and other high-tech capital is *less than* the contribution of outsourcing. In contrast, when computers are measured by their share of investment (and the high-technology capital share is also included), we see that these variables account for 31% of the shift toward nonproduction labor, which *exceeds* the contribution of outsourcing. Thus, whether outsourcing is more or less important than computers depends of whether the latter are measured as a share of the capital stock, or as a share of investment. It is fair to conclude that both these variables are important explanations of the shift towards nonproduction labor, with their exact magnitudes depending on how they are measured.

In related work, Morrison Paul and Siegel (2000) find a negative correlation between the demand for less-educated labor and high-tech capital, R&D investment, imports (as a share of output), and service purchases (i.e., domestic outsourcing), and a positive correlation between the demand for more-educated labor and high-tech capital, R&D investment, and imports. Their estimated impact of high-tech capital and R&D investment on skill upgrading is greater than that for imports or domestic outsourcing of services.

From these results, it appears that foreign outsourcing from US manufacturing is associated with the increased relative demand for skilled labor, as predicted by the theory outlined in section 3. One important question is what form this outsourcing takes. It may occur through foreign direction investment (FDI), as multinationals move production of parts and components or product assembly abroad, or it may occur through a shift in contracting practices,

in which firms replace domestic production of intermediate inputs with imports purchased from arms-length suppliers located abroad. The first case appears to be consistent with what has occurred in the US automobile and semiconductor industries, while the second case appears to be consistent with what has occurred in the US footwear and personal-computer industries.

Slaughter (2000b) finds that during the 1980's FDI was *not* an important channel for moving US jobs abroad or for skill upgrading at home. Between 1977 and 1989, employment in majority-owned affiliates of US multinational enterprises (MNEs) fell, as it did in the US parents of these plants. Similar to US manufacturing plants, affiliates of US MNEs spent the 1980's shifting employment towards nonproduction workers and raising the capital intensity of production. Slaughter estimates a specification similar to (15), in which he includes as regressors shipments, the capital/shipments ratio, and the ratio of economic activity in foreign affiliates of US MNEs to economic activity in US manufacturing plants. The last variable captures the extent to which US MNEs are transferring production abroad. For several measures of economic activity in foreign affiliates, there is a weak *negative* correlation between the change in the nonproduction wage share and the change in foreign production transfer by US MNEs. Combined with the results in Feenstra and Hanson (1999), these results suggest that FDI is not the means through which outsourcing has induced skill upgrading in US manufacturing.¹⁹ In related work, Blonigen and Slaughter (1999) find that inward FDI in US manufacturing during the 1980's, such as the construction of Japanese auto plants and other facilities, also did not contribute to skill upgrading in US manufacturing industries.²⁰

The United States is by no means the only country that engages in outsourcing. Many European nations are outsourcing, as well as industrialized Asian countries including Japan, Hong Kong, Korea and Taiwan (Ng and Yeats, 1999). In one study for the United Kingdom,

Anderton and Brenton (1997) measure outsourcing by imports from *low-wage* countries. They find that such imports can account for about 40% of the rise in the wage bill of skilled workers of 1970-83, and approximately one-third of their increase in relative employment.

Over the last several decades, Japan has steadily moved production activities abroad, investing in both low-income and high-income countries. For firms in the Japanese electronics industry, overseas employment now greatly exceeds employment in Japan. These production shifts have coincided with skill upgrading at home. Since the early 1970's Japanese manufacturing industries have had a steady increase in the relative employment and relative total compensation of more-educated workers. Applying a specification similar to (15) to data on Japanese manufacturing plants over the period 1965-1990, Head and Ries (2001) find a strong *positive* correlation between the change in a firm's nonproduction wage share and the change in a firm's share of employment in low-income countries. This suggests that as Japanese manufacturing firms move production to *low-wage* countries they raise the skill intensity and the demand for skilled labor at home. The correlation between the nonproduction wage share and the employment share in high-income countries, in contrast, is statistically insignificant in most cases.²¹

During the 1960's and 1970's, Hong Kong produced and exported labor-intensive manufactures, such as apparel, textiles, footwear, toys, and consumer electronics. Since China began to open its economy to foreign trade and investment in the late 1970's, Hong Kong has begun to specialize in business services related to trade and investment in China. Over the last two decades, many Hong Kong manufacturing firms have moved their production facilities to China, and to the neighboring province of Guandong in particular, which they manage from headquarters in Hong Kong (Sung, 1997). Hong Kong firms typically supply plants in China

with raw materials and often ship the goods through Hong Kong for final processing before exporting them to a final destination. As Hong Kong has shifted production to China, manufacturing has become a less important part of the Hong Kong economy, declining from 24% of GDP in 1980 to 7% of GDP in 1996 (Enright et al., 1997).

Hsieh and Woo (1999) find that outsourcing from Hong Kong to China has been associated with an increase in the relative demand for skilled labor in Hong Kong. Between 1981 and 1996, both the relative wage and relative supply of more-educated workers rose in Hong Kong, which suggests that there was an increase in the relative demand for skilled labor. The extent of skill upgrading in Hong Kong manufacturing was dramatic. The share of nonproduction workers in Hong Kong manufacturing employment rose from 13.1% in 1976 to 47.0% in 1996. Outsourcing to China appears to be an important contributing factor to this employment shift. Using a regression similar to (15), Hsieh and Woo find a strong positive correlation between the nonproduction wage share and imports from China (expressed as the ratio of industry imports from China to either industry absorption or total materials purchases) for Hong Kong industries. Over the period 1976-1996, outsourcing to China can account for 45-60% of the increase in the nonproduction wage share in Hong Kong manufacturing.

Outsourcing may also raise the relative demand for skilled labor in the country to which production is transferred. In Feenstra and Hanson (1996), firms in the skill-abundant North use firms in the nonskill-abundant South to produce intermediate inputs. Assuming wages differ between nations, the North specializes in high-skill tasks and the South specializes in low-skill tasks. If Northern firms outsource production to the South, they will choose to move the least skill-intensive activities that they perform. By moving these activities to the South, the average skill intensity of production *rises* in the North. The same also happens in the South, since the

South initially specializes in the most low-skill tasks. Thus, outsourcing from the North to the South raises the relative demand and so the relative earnings of high-skilled workers in *both* countries, contributing to a global increase in wage inequality.

During the 1980's Mexico, liberalized foreign investment and trade and experienced an increase in the relative wage of skilled labor (Hanson and Harrison, 1999). In the period following reform, FDI in Mexico was concentrated in foreign assembly plants, known as *maquiladoras*, most of which are located in Mexican states along the US border. These assembly plants are created, in most cases, by US firms outsourcing unskilled-labor intensive production activities to Mexico. Feenstra and Hanson (1997) find that for the period 1975-1988 the shift in Mexican manufacturing towards foreign assembly plants can account for 45% of the observed increase in the country's nonproduction wage share.²²

Outsourcing changes the relative wage by inducing an outward shift in the relative demand for skilled labor. An alternative view is that international trade changes factor prices by flattening labor demand curves, making them more elastic. Leamer (1998) presents an extreme version of this story, in which the transition of an economy from autarky to trade transforms an economy's labor demand curve from being downward sloping to being horizontal, at least over sections that correspond to diversified production. Extending this logic, Rodrik (1997) identifies several mechanism through which greater economic integration between countries may make labor demand curves flatter. In one of the few attempts to test this hypothesis, Slaughter (2001) estimates the own-price elasticity of labor demand for production and nonproduction workers in two-digit US manufacturing industries over the period 1960-1991. Over the entire sample, demand became more elastic for production labor, but not for nonproduction labor. The sectors

with the largest increase in elasticities were food and tobacco, apparel and textiles, wood and paper, and primary and fabricated metals, which include some of the least-skill intensive manufacturing industries. The demand for production labor became more elastic in industries with more outsourcing, more investment in computers, and more investment in high-tech capital overall. These results are robust to controls for industry fixed effects but not time fixed effects, suggesting that changes in labor-demand elasticities are dominated by a trend component.

Estimation of Zero-Profit Conditions

In the second empirical method, we drop the short-run framework that was used above, and instead suppose that capital can be adjusted along with skilled and unskilled labor. The industry cost function is then written as in (6), which is re-expressed as:

$$C_m(w_m, q_m, r_m, Y_m, p, p_m) \equiv \min_{L_m, H_m, K_m} w_m L_m + q_m H_m + r_m K_m, \text{ subject to (4)}. \quad (16)$$

Notice that in (16) we have allowed the factor prices w_m , q_m , and r_m to *differ* across the industries $m=1, \dots, M$. This reflects the empirical fact that factor prices, and wages in particular, do differ across industries (Krueger and Summers, 1988); this will turn out to be important in what follows. As before, the relative price of imported inputs enters this cost function because it also appears in the production function (4); we will replace this by the vector z_m , which includes other structural variables. Since the production function (4) is linearly homogeneous in inputs, then we can rewrite the cost function in (16) as,

$$C_m(w_m, q_m, r_m, Y_m, Z_m) = Y_m c_m(w_m, q_m, r_m, Z_m), \quad (16')$$

where $c_m(w_m, q_m, r_m, Z_m)$ denotes the unit-cost function.

The zero-profit conditions in the industries are therefore expressed as,

$$p_m = c_m(w_m, q_m, r_m, z_m), \quad m=1, \dots, M. \quad (17)$$

In our theoretical model of section 3, we examined how changes in product prices would affect factor prices. Now, however, the presence of the structural variables z_m mean that the changes in prices reflect more than just changes in factor prices. Indeed, taking the difference between these, we can define *total factor productivity* (TFP) in the industry as,

$$TFP_m \equiv (\theta_{mL} \hat{w}_m + \theta_{mH} \hat{q}_m + \theta_{mK} \hat{r}_m) - \hat{p}_m, \quad (18)$$

where the cost-shares of the three factors sum to unity, $\theta_{mL} + \theta_{mH} + \theta_{mK} = 1$. Productivity improvements mean that factor prices can rise more than product prices (or conversely, that product prices can fall further). Note that (18) is the “dual” definition of productivity, and empirically it is very close to the “primal” definition, which is the growth in output minus a weighted average of the growth in inputs. In either case, we should think of changes in the structural variables z_m as the underlying cause of changes in productivity.²³

Shuffling the terms in (18) slightly, and replacing the instantaneous change in prices with discrete changes like $\Delta \ln p_m$, we obtain the equation,

$$\Delta \ln p_m = -TFP_m + \theta_{mL} \Delta \ln w_m + \theta_{mH} \Delta \ln q_m + \theta_{mK} \Delta \ln r_m, \quad m=1, \dots, M. \quad (19)$$

We consider estimating (19) as a linear regression across industries, where the *data* are the

change in log prices, productivity, and the factor cost-shares, while the change in factor-prices are *estimated* as regression coefficients. That is, we estimate the *implied* change in factor-prices ω_L , ω_H , and ω_K from the regression:

$$\Delta \ln p_m = -TFP_m + \theta_{mL}\omega_L + \theta_{mH}\omega_H + \theta_{mK}\omega_K + \varepsilon_m, \quad m=1, \dots, M. \quad (20)$$

where ε_m is an error term, specified more fully below. We interpret the coefficients ω_L , ω_H , and ω_K as the change in factor prices that are *mandated by* the change in product prices, which is the dependent variable in (19). Baldwin and Hilton (1984) were among the first to estimate a price regression like (20). Recent applications include Sachs and Shatz (1994), Baldwin and Cain (1997), Krueger (1997), Leamer (1998), Feenstra and Hanson (1999), and Slaughter (2000b).

Estimates of (20) for 447 US manufacturing industries, over 1979-1990, are provided in Table 5. The dependent variable is the log change in the industry output price over the period, divided by the number of sample years to obtain an annualized difference. We use the primal measure of TFP, expressed as an annualized difference. The other independent variables are the average cost-shares (over the first and last year for the period) for production labor, nonproduction labor, and capital; the materials cost share times the log change in materials prices; and the energy cost share times the log change in energy prices.

In columns (1) and (2), we constrain the coefficients on the materials share times the materials price and the energy share times the energy price to be unity, which transforms the dependent variable to be the log change in value-added prices. This approximates the specification in Leamer (1998). In column (1), the coefficients on the labor shares imply a *decrease* in the nonproduction-production wage gap, since the nonproduction-production relative wages is mandated to change by $2.30\% - 3.06\% = -0.76\%$ per year, which is consistent with the

results in Leamer (1998). In fact, the nonproduction-production wage gap *rose* by 0.74% per year (actual annual average changes in factor prices are shown in column (5) of Table 5). In column (2), we follow Sachs and Shatz (1994) and drop the office-equipment industry (SIC 3573), which *reverses* the predicted change in wage inequality. Now, nonproduction wages are mandated to rise by 1.5% per year more than production wages. In column (3), we approximate Krueger's (1997) specification by dropping TFP as a regressor, while estimating coefficients on materials and energy. There is again a mandated rise in the nonproduction-production wage gap, but one that is much larger than the actual increase in relative wages.

The estimates in Table 5 are troubling because they show that slight changes in the data, such as dropping the office-equipment industry, have a dramatic effects on the results. While it is true that office equipment is an outlier, the sensitivity of the results to the specification suggests that something more basic is going on. To address this, let us ask: why do the estimates of ω_L , ω_H , and ω_K from (20) differ at all from the *actual average* change in manufacturing wages (shown in column (5)), which we denote by $\overline{\Delta \ln w}$, $\overline{\Delta \ln q}$, and $\overline{\Delta \ln r}$? The overbar indicates that we are averaging the change in factor-prices over all manufacturing industries. By just comparing (19) and (20), it seems that there should be some close connection between the estimates ω_L , ω_H , and ω_K and these average actual factor price changes, but we need to uncover what this connection is.

To achieve this, let us make the transition from (19) to an estimating equation more carefully. First, notice that we can rewrite (19) as,

$$\Delta \ln p_m = -TFP_m + \theta_{mL} \overline{\Delta \ln w} + \theta_{mH} \overline{\Delta \ln q} + \theta_{mK} \overline{\Delta \ln r} + \varepsilon_m \quad (19')$$

where,

$$\varepsilon_m \equiv \theta_{mL}(\overline{\Delta \ln w} - \Delta \ln w_m) + \theta_{mH}(\overline{\Delta \ln q} - \Delta \ln q_m) + \theta_{mK}(\overline{\Delta \ln r} - \Delta \ln r_m). \quad (20')$$

Thus, we replace the *industry* wage changes on the right of (19) by the *average* wage changes, and incorporate the difference between these two into an error term. In economic terms, ε_m reflects interindustry wage differentials – i.e., the difference between wages paid in each industry and the manufacturing average. It is well known that these wage differentials vary systematically across industries (with capital-intensive industries paying higher wages), and are fairly stable over time (Krueger and Summers, 1988).

We can estimate (19') as the regression (20), where now we have been careful to derive the error term ε_m . But this derivation is enough to answer the question we posed above: the estimates of ω_L , ω_H , and ω_K obtained from (20) will be unbiased estimates of the *average actual* factor price-changes in (19') if and only if the error term ε_m in (20') is *uncorrelated* with the cost-shares θ_{mL} , θ_{mH} , θ_{mK} . This result follows directly from the properties of ordinary least-squares, whereby the independent variables need to be uncorrelated with the error term to obtain unbiased estimates. But this property is unlikely to be true in our data. Industries such as office equipment have both a high share of nonproduction labor (e.g., engineers), and probably the fastest growing industry wage differential, as these workers have had very rapid wage gains. This means that the error term ε_m is negative for office equipment, resulting in a negative correlation with the cost-share of nonproduction labor. Indeed, this negative correlation may explain why the *estimated* change in nonproduction wages is *lower* in column (1) than in column (2) of Table 5 (and why both estimates are lower than the actual change in nonproduction wages in column (5)).

To correct this problem, we can simply include the error term ε_m as an additional regressor in the equation, reflecting the change in interindustry wage differentials. It is convenient to combine ε_m with TFP_m , obtaining a measure of “effective” TFP:

$$\begin{aligned} ETFP_m &\equiv TFP_m - \varepsilon_m \\ &= (\theta_{mL} \overline{\Delta \ln w} + \theta_{mH} \overline{\Delta \ln q} + \theta_{mK} \overline{\Delta \ln r}) - \Delta \ln p_m \quad . \end{aligned} \quad (21)$$

Thus, this measure of effective productivity shows how the *average manufacturing factor-price changes*, weighted using the cost-share in each industry, differ from the change in product price of that industry. Making use of (21), the regression in (20) is written once again as,

$$\Delta \ln p_m = -ETFP_m + \theta_{mL} \omega_L + \theta_{mH} \omega_H + \theta_{mK} \omega_K, \quad m=1, \dots, M. \quad (22)$$

Now, there is no error term at all in this regression, so it ought to provide a perfect fit when estimated. This will not be true exactly in our data, since we are using the primal rather than the dual measure of TFP to construct effective TFP in (21). These priors are confirmed in columns (4) and (5) of Table 5. In column (4), the dependent variable is the log change in value-added prices; in column (5), it is the log change in output prices, since we allow the coefficients on the materials and energy shares to differ from unity. In either specification, the coefficients on the labor and capital shares are extremely close to the actual average annual percentage changes in factor prices, which are 4.71 for production labor, 5.44 for nonproduction labor, and 3.95 for capital. Thus, when we properly estimate (20), we end up with an identity!

Summarizing our results so far, we started with the goal of estimating the zero-profit conditions, to obtain “mandated” changes in factor prices that are consistent with the change in product prices. A number of researchers have estimated an equation like (20), without much

attention to the error term in this regression. When we carefully derive the error term, as in (20'), we then realize that it may well be correlated with the factor cost-shares, which are the independent variables. To correct for this we can include the error term as data, by incorporating it into “effective” total factor productivity, as in (22). But now we encounter another problem: this gives essentially a perfect fit, just reproducing the *actual* change in factor prices. That means the regression does not provide us with any new information at all! This is a discouraging finding, and calls into question the whole approach.

To make further progress, Feenstra and Hanson (1999) propose a two-step estimation procedure. In the first step, we combine the variables $\Delta \ln p_m + \text{ETFP}_m$ that appear in (22), and regress these on the structural variables z_m . Supposing that there are only two structural variables, z_{1m} and z_{2m} , we therefore run the regression:

$$\Delta \ln p_m + \text{ETFP}_m = \eta_0 + \eta_1 \Delta z_{1m} + \eta_2 \Delta z_{2m} \quad , \quad m=1, \dots, M. \quad (23)$$

We then take the estimated coefficients $\hat{\eta}_1$ and $\hat{\eta}_2$, and use these to construct the dependent variables for the second-step regressions,

$$\hat{\eta}_1 \Delta z_{1m} = \theta_{mL} \omega_{1L} + \theta_{mH} \omega_{1H} + \theta_{mK} \omega_{1K} \quad , \quad (24a)$$

$$\hat{\eta}_2 \Delta z_{2m} = \theta_{mL} \omega_{2L} + \theta_{mH} \omega_{2H} + \theta_{mK} \omega_{2K} \quad , \quad m=1, \dots, M. \quad (24a)$$

That is, we use the estimated coefficients $\hat{\eta}_1$ and $\hat{\eta}_2$ times each structural variable as the dependent variables in (24), and regress these on the factor cost-shares. The coefficients obtained from the second-stage regression, ω_{1L} , ω_{1H} , ω_{1K} and ω_{2L} , ω_{2H} , ω_{2K} , are interpreted as *the portion of the total change in factor-prices that are explained by that structural variable*. In this

way, we are taking the total change in factor-prices, and decomposing it into parts that are explained by each structural variable.

In the estimation of (23) for US manufacturing industries over the period 1979-1990, Feenstra and Hanson (1999) find positive and statistically significant correlations between TFP-adjusted value-added prices and foreign outsourcing, the computer share of the capital stock, and the computer share of investment (but not with the high-tech capital share). This is consistent with these structural variables having positive effects on productivity, as expected, *and* on value-added prices.²⁴ The latter effect arises from the non-neutral impact that the structural variables have on industry productivity – outsourcing and capital upgrading both induce shifts away from production labor and towards nonproduction labor – which then leads to changes in relative product prices (in particular, raising value-added prices for skill-intensive goods). Over the period 1979-1990, the structural variables account for 11% to 23% of the variation in TFP-adjusted value-added prices in US manufacturing.

The second stage of this technique is to decompose the change in TFP-adjusted value-added prices into portions attributable to each structural variable and then, following (24), regress these decomposed product-price changes on the factor-cost shares to obtain mandated changes in factor prices. The results suggest that both outsourcing and capital upgrading contributed to rising wage inequality in the 1980's. Over the 1979-1990 period, outsourcing accounts for 15% of the increase in the relative wage of nonproduction workers, and computers measured using *ex post* rental prices account for 35% of this increase; thus, computers are twice as important as outsourcing. When instead the computer share of the capital stock is measured using *ex ante* rental prices, then outsourcing explains about 25% while computers explain about 20% of the increase in the nonproduction/production wage. Finally, when the computer share of the capital

stock is replaced with the computer share of *investment*, then the contribution of outsourcing falls to about 10%, while the contribution of computers rises so much that it explains the *entire* increase in the relative wage. Thus, as in Table 4 where we examined the change in the nonproduction labor *share*, when we now consider the factors influencing the *relative wage*, we find that both outsourcing and computer expenditure are important with their exact magnitudes depending on how these variables are measured.

Economic integration between countries may, of course, also contribute directly to changes in factor prices, as lower tariffs or transportation costs lead to changes in product prices which would then affect factor prices in standard Stolper-Samuelson (1941) fashion. In principle, one could uncover the impact of changing tariffs and transportation costs on product prices and productivity, following (23), and use these results to estimate their impact on factor prices, following (24). Haskel and Slaughter (2000) show that in the 1970's US manufacturing industries with higher tariffs and higher transportation costs tended to have lower relative employment of nonproduction workers, and that over the period 1974-1988 reductions in tariffs and transportation costs were larger in less-skill-intensive industries. This is suggestive evidence that falling trade costs may have contributed to rising wage inequality. But other evidence is missing: changes in tariffs and transportation costs do poorly in explaining changes in product prices across US industries over the 1974-1988 period. Robertson (2000) finds stronger results for Mexico: over 1987-1993, when Mexico's skilled-unskilled wage gap rose, tariffs fell more in less-skill-intensive industries, and over 1993-1998, when Mexico's skilled-unskilled wage gap fell, tariffs fell more in skill-intensive sectors.

The United Kingdom is another country that has experienced rising wage inequality of the same magnitude as the United States. Haskel and Slaughter (2001) apply the two-stage

estimation procedure of Feenstra and Hanson (1999) to data on UK manufacturing industries over the period 1960-1990, using as structural variables union density (the share of union workers in industry employment), industry concentration (share of sales by the five largest firms), innovations per industry, import prices, and computerization (share of firms in the industry using computers). They find that TFP growth is higher in industries with more innovations, lower initial union density, lower initial sales concentration, and larger reductions in import prices (but is unrelated with computerization). Product price changes are lower in industries with smaller changes in import prices. During the 1980's, when UK wage inequality rose, the structural variable that appears to have contributed most to the increase in the skilled-unskilled wage gap is industry innovation. The contribution of import prices is comparatively small. This contrasts with research (Anderton and Brenton, 1997) showing that rising imports over 1970-83 is a significant determinant of the nonproduction labor *share* in the U.K.

Estimation of Economy-wide GDP Function

The third empirical method we shall discuss takes the longest-run view of the economy, in which capital and all other factors fully adjust to their equilibrium levels across industries. Rather than focusing on each industry, we now look at the economy overall, and consider how gross domestic product (GDP) is produced from the total factor endowments, given the prices of domestic and traded goods. Letting (L_m, H_m, K_m) denote the factor demands within each industry $m=1, \dots, M$, the sum of these cannot exceed the endowments (L, H, K) :

$$\sum_{m=1}^M L_m \leq L, \quad \sum_{m=1}^M H_m \leq H, \quad \text{and} \quad \sum_{m=1}^M K_m \leq K . \quad (25)$$

Using the production functions in (3) for each industry, the GDP function for the economy is defined as:

$$G(L, H, K, P) \equiv \max_{L_m, H_m, K_m} \sum_{m=1}^M F_m(L_m, H_m, K_m, p_m, p), \quad \text{subject to (25),} \quad (26)$$

where $P=(p_1, \dots, p_M, p)$ denotes the vector of all product prices p_m as well the prices p of all imported intermediate imports. Let us say there are $N > M$ prices in total, so that $P=(p_1, \dots, p_N)$. Within the list of M industries, we are including both manufacturing and services, whether these products are traded or not.

In order to estimate (26) we need to choose a functional form, and as in our discussion of the cost function, a convenient choice is the translog function. This is written in a general notation as,

$$\begin{aligned} \ln G = & \alpha_0 + \sum_{m=1}^N \alpha_m \ln p_m + \sum_{k=1}^K \beta_k \ln v_k + \frac{1}{2} \sum_{m=1}^N \sum_{n=1}^N \gamma_{mn} \ln p_m \ln p_n \\ & + \frac{1}{2} \sum_{k=1}^K \sum_{\ell=1}^K \delta_{k\ell} \ln v_k \ln v_\ell + \sum_{m=1}^N \sum_{k=1}^K \phi_{mk} \ln p_m \ln v_k \end{aligned} \quad (27)$$

where p_m denotes the prices of the outputs and imported inputs, $m=1, \dots, N$, while v_k denotes the endowments of the factors of production, $k=1, \dots, K$. For the GDP function in (26), there are just three endowments – skilled and unskilled labor, and capital. In order to ensure that the GDP function is linearly homogeneous in prices, we impose the requirements,²⁵

$$\sum_{n=1}^N \alpha_n = 1 \quad \text{and} \quad \sum_{m=1}^N \gamma_{mn} = \sum_{n=1}^N \gamma_{mn} = 0. \quad (28)$$

As usual, we can differentiate the GDP function with respect to factor endowments to obtain factor prices, $\partial G/\partial v_k = w_k$. It follows that the derivative of the log of GDP with respect to the log of endowments, $\partial \ln G/\partial \ln v_k = w_k v_k/G$, equals the share of GDP paid to each factor, which we denote by s_k , $k=1, \dots, K$. Computing this for the GDP function in (27), we obtain:

$$s_k = \beta_k + \sum_{\ell=1}^K \delta_{k\ell} \ln v_\ell + \sum_{m=1}^N \phi_{mk} \ln p_m, \quad k=1, \dots, K. \quad (29)$$

In addition, differentiating the GDP function with respect to a product price yields output of that good (inclusive of exports), and differentiating with respect to an import price yields the negative of imports. Therefore, the derivative of the log of GDP with respect to the log of a price equals the share of GDP obtained from that output, which we denote by s_n , $n=1, \dots, N$. Note that for imported inputs, these shares are *negative*. Computing these from (27):

$$s_m = \alpha_m + \sum_{n=1}^N \gamma_{mn} \ln p_n + \sum_{k=1}^K \phi_{mk} \ln v_k, \quad m=1, \dots, N. \quad (30)$$

Thus, given annual data on product prices and quantities, along with factor prices and endowments, we can estimate (29) and (30) as a system of linear equations. This will allow us to determine the effect of product prices on factor shares – measured by ϕ_{mk} . Notice that these coefficients enter both the factor-share equations in (29), and the product-share equations in (30), so that estimating these simultaneously allows us to test the restriction that ϕ_{mk} estimated from (29) and (30) are equal. If this hypothesis is accepted, then the restriction can be imposed, and this allows the estimates of ϕ_{mk} to become more precise. An advantage of this system approach is that the number of years (say T) is multiplied by the number of equations, $(N-1)$ for products

plus $(K-1)$ for factors, to obtain the total number of observations $T(N+K-2)$.²⁶ Thus, even though the number of parameters being estimated is large, we can obtain reasonable estimates even if we only have annual data for one or two decades.

Estimates of the GDP function have been made by a number of authors for various countries, including the United States. The reader is referred to Kohli (1991) for the most comprehensive treatment. For the purposes of this survey, we are interested in applications that distinguish skilled versus unskilled labor. Tombazos (1999) attempts to make this distinction by identifying *industries* that are intensive in skilled or unskilled labor, and then forming aggregate wages and employment indexes over each group of industries; these indexes are then used as a proxy for the price and quantity of skilled and unskilled labor. Imports are treated as an input into the production process, just as they are in the GDP function (26). Tombazos incorporates skilled labor, unskilled labor, capital and imports into the estimation of an aggregate cost function for the U.S., over 1967-1994, with a single aggregate output (including exports). His major conclusion is that a drop in the import price reduces the demand for unskilled labor, but *raises* the demand for skilled labor in the US. This is highly consistent our theoretical model of section 3. According to Tombazos, an effect like this holds for the US economy in the aggregate. Missing from his analysis, though, is a discussion of *how much* import prices have fallen, and therefore, how much of the shift towards skilled labor can be explained by this channel of influence.

This shortcoming does not appear in the work of Harrigan and Balaban (1999) and Harrigan (2000). Harrigan and Balaban estimate the system of equations in (29) and (30) for the United States over the period 1963-1991 using data on four factors (high-school dropouts, high-school graduates, college graduates, capital), and four sectors (skill-intensive traded goods,

unskilled-intensive traded goods, skill-intensive nontraded goods, and unskilled-intensive nontraded goods). Thus, imports are not explicitly identified. In contrast, Harrigan (2000) has two categories of outputs (skill-intensive and unskilled-intensive final output), and seven factors including imports (oil imports, two other groups of imports, and the three types of labor and capital). It turns out that changes in the import prices have been quite small in comparison with other prices changes, especially in nontraded goods, so that changes in import prices *are not* an important explanation for changes in wages. We therefore focus below on the results of Harrigan and Balaban, which except for imports, are similar to those of Harrigan.

With the estimated coefficients from the share equations (29) and (30) in hand, Harrigan and Balaban calculate wage elasticities with respect to factor quantities and product prices. As expected, the own-quantity elasticity of each factor price is negative. Increasing the supply of capital raise wages for all workers, but these elasticities are increasing in education levels, such that a 10% increase in the capital stock would increase the college/high-school graduate relative wage by about 3.5%, and the college/high-school dropout relative wage by about 8%. The wage elasticities of *traded* goods prices are imprecisely estimated, while those for *nontraded* goods are somewhat surprising. Increases in prices of skill-intensive nontraded goods raises wages for college graduates and high-school dropouts, but lowers wages for high-school graduates, and increases in prices of unskilled-intensive nontraded goods has a large positive effect on high-school-graduate wages, a moderate positive effect on college wages, and a negative effect on high-school-dropout wages.

Putting the estimated wage elasticities together with observed changes in factor supplies and product prices, we can decompose the contribution of different variables to the observed change in factor prices. While capital accumulation contributed to an increase in the relative

wage of college graduates, this effect was largely offset by increases in the supply of college graduates. The big changes during the latter part of the sample period were an *increase in the relative price of skill-intensive nontraded goods*, such as finance, insurance and real estate. This had the largest impact on *raising* the college/high-school-graduate relative wage. Conversely, there was a *decrease in the relative price of unskilled-intensive nontraded goods*, such as wholesale and retail trade, which had the largest impact on *reducing* the relative wage of high-school dropouts. In short, the increase in the relative wage of skilled labor, in the 1980s and beyond, is highly correlated with the rise in the price of nontraded goods that use skilled labor, and similarly for unskilled labor where both the relative wage and price fell.

These correlations beg the questions of whether the change in nontraded prices caused the change in wages, or conversely. We provided a theoretical example in section 3 of a case where capital leaves the country, increasing the relative wage of skilled labor, and therefore raising the price of skill-intensive nontradables and lowering the price of unskilled-intensive nontradables. This story would be consistent with the estimates of Harrigan and Balaban. We cannot rule out, however, the idea that the nontradables prices are changing due to some other reason (e.g. rising incomes and demand leading to an increase in the price of skill-intensive nontradables), which is therefore the proximate cause of the change in wages. As Harrigan (2000, p. 186) puts it: “To my knowledge, there are no scholarly studies of relative price determination in the United States that might shed light on the causes of the changes shown...and until we understand the cause of these price changes we cannot rule out an important role for import competition.”

5. Conclusions

There is an emerging view in the literature on wage inequality in the United States and other advanced countries that technological change matters for changes in the wage structure but international trade does not (Katz and Autor, 1999). The research we survey in this chapter fails to support this conclusion. While there is abundant evidence of skill-biased technological change, it also appears that international trade, in the form of foreign outsourcing, contributes to skill upgrading and increases in the skilled-unskilled wage gap.

The argument against trade is based, in part, on a misreading of the data. Stable trade to GDP ratios, an apparent increase in the relative price of skill-intensive goods, and employment shifts towards skilled workers that occur mainly within, rather than between, industries are all cited as evidence that trade cannot have contributed to rising wage inequality. This line of reasoning emphasizes trade in final goods and ignores the globalization of production and recent dramatic increases in trade in intermediate inputs. Much recent growth in trade has resulted from firms breaking industries apart by locating low-skill activities in low-wage countries and high-skill activities in high-wage countries. When trade takes this form, its impact on relative prices and factor allocation can be much different than that predicted by standard models of trade in final goods. Recent literature shows that trade to merchandise GDP ratios have risen sharply in recent years, with much of the growth in trade attributable to intermediate inputs, that changes in the relative prices of domestic versus imported goods are consistent with trade shifting out the relative demand for skilled labor, and that trade in intermediate inputs is consistent with skill upgrading being a within-industry phenomenon.

Beyond the *prima facie* case that trade raises wage inequality, there is evidence of a direct link. Using data on changes in industry behavior over time, we see that foreign outsourcing is

associated with increases in the share of wages paid to skilled workers in the United States, Japan, Hong Kong, and Mexico. In several of these cases, outsourcing can account for half or more of the observed skill upgrading. For the United States, there is evidence that during the 1980's and 1990's outsourcing contributed to changes in industry productivity and product prices that in turn mandated increases in the relative wage of skilled labor.

Existing literature has just begun to scratch the surface of how the globalization of production changes industry structure and factor demand in advanced and emerging economies. There is as of yet little research on foreign outsourcing in Eastern Europe, Southeast Asia (Cheng and Kierzkowski, 2001), or Central America and the Caribbean (Feenstra, Hanson and Swenson, 2000), though anecdotal evidence suggests that it is an important mechanism through which countries in these regions integrate themselves into the world economy. There are clearly rich opportunities for research in this area.

Global production sharing and trade in intermediate inputs matter more generally for how we apply trade models to data. In tests of the Heckscher-Ohlin model, it is standard to assume that exports are produced entirely by combining domestic factors of production with domestically-produced intermediate inputs. We now know that this assumption is wrong. Antweiler and Trefler (2001), Davis and Weinstein (2000) and Trefler and Zhu (2000) show that trade in intermediate inputs can help resolve the mystery of the missing trade (Trefler, 1995), in which the estimated factor content of trade is near zero, and that existing tests of trade theory may produce severely biased estimates of the factor content of trade if they fail to account for global production. While we can easily address this problem by adopting more general trade models in which trade occurs in both final and intermediate goods, the literature has yet to take this necessary step.

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Endnotes

¹ Alternatively referred to as outsourcing (Katz and Murphy, 1992, Feenstra and Hanson, 1996), de-localization (Leamer, 1998), fragmentation (Deardorff, 2001; Jones and Keirzkowski, 1997, Arndt and Kierzkowski, 2000), intra-product specialization (Arndt, 1997 and 1998), intra-mediate trade (Antweiler and Trefler, 1997), vertical specialization (Hummels, Ishii, and Yi, 2001), and slicing the value chain (Krugman, 1995), this phenomena refers to the geographic separation of activities involved in producing a good (or service) across two or more countries. The term “production sharing” was coined by management consultant Peter Drucker (“The Rise of Production Sharing,” *The Wall Street Journal*, March 15, 1977), as we have adopted in our title.

² For a detailed discussion, see Katz and Autor (1999), whose wage figures we report below.

³ Only the highly skilled have had large real-wage gains. For the 1979-1995 period, real wages for those with 18 or more years of education rose by 14.0% and for those with 16 to 17 years of education rose by only 1.0%.

⁴ The breakdown of workers according to whether or not they are engaged in production activity is made in the U.S. *Annual Survey of Manufactures*, and is used as a proxy for the occupational-class or skill-level of workers. While there are problems with using the production-nonproduction classification as a proxy for skill, there is evidence that in practice the classification shows similar trends as using skill categories (Berman, Bound and Griliches, 1994; Berman, Bound and Machin, 1998; Sachs and Shatz, 1994). The increase in the nonproduction-production relative wage is only a small part of the total increase in wage inequality between

more- and less-skilled workers that occurred during the 1980s. See Katz and Murphy (1992) and Katz and Autor (1999) for a discussion.

⁵ Rising wage inequality has been relatively modest in Australia, Canada, Japan, and Sweden. Among advanced countries, only the United Kingdom has had relative wage changes comparable to the United States (Machin, 1996).

⁶ See the surveys by Freeman (1995), Johnson and Stafford (1999), Richardson (1995), Wood (1995), and Feenstra (1998) and the volumes by Bhagwati and Kesters (1994) and Collins (1998).

⁷ Data on the ratio of trade to GDP is provided in Feenstra (1998), who also computes the ratios of merchandise trade to merchandise value-added, as discussed below.

⁸ Krugman (1995, 2000) provide the *theoretical* argument for why a small share of trade in the U.S. makes it unlikely that trade can account for the change in wages. For an alternative views, see Deardorff (2000), Leamer (2000), and Panagariya (2000).

⁹ Imported intermediate inputs have also been computed for nine OECD countries by Hummels, Ishii, and Yi (2001). They find that 30% of OECD exports are attributable to imported intermediate inputs used in their production.

¹⁰ See Slaughter (2000a) for a discussion of literature on relative-price changes.

¹¹ For a discussion of the Mexican *maquiladora*, see Feenstra and Hanson (1997). The export of inputs from the U.S. for assembly overseas also occurs in other countries under the U.S. “offshore assembly” program: see Feenstra, Hanson and Swenson (2000).

¹² Other models that emphasize the movement of production activities across countries in a product-cycle framework, and the effect on wages, include Dinopoulos and Segerstrom (1999),

Glass and Saggi (2001) and Zhu (2000a), with corresponding empirical work by Zhu (2000b). The reader is also referred to the collection of papers in the June 2001 issue of the *Journal of International Economics*, 54(1).

¹³ To verify this, use $\hat{p} = 0$ in (9) and then solve for \hat{w} and \hat{q} as: $\hat{w} = -\hat{r} (\theta_{2H}\theta_{1K} - \theta_{1H}\theta_{2K})/(\theta_{1L}\theta_{2H} - \theta_{1H}\theta_{2L})$, and $\hat{q} = \hat{r} (\theta_{2L}\theta_{1K} - \theta_{1L}\theta_{2K})/(\theta_{1L}\theta_{2H} - \theta_{1H}\theta_{2L})$. Combining these, we readily obtain $\hat{w} - \hat{q} = -\hat{r} (\theta_{1K} - \theta_{2K})/(\theta_{1L}\theta_{2H} - \theta_{1H}\theta_{2L})$, which is negative when $\hat{r} > 0$ under our assumptions that activity 1 is intensive in both unskilled labor and capital.

¹⁴ Sachs and Shatz (1998, pp. 220-221) emphasize that this capital outflow *need not* be foreign direct investment, in which a domestic firm takes ownership of capital abroad. Rather, the domestic capital could simply be sold to a foreign firm (which would not show up as foreign direct investment).

¹⁵ Without loss of generality, we also impose the symmetry restrictions $\gamma_{ij} = \gamma_{ji}$ and $\delta_{jk} = \delta_{kj}$.

¹⁶ Generally, the dependent variables in the system (14) sum to unity, which means that one of the equations can be derived from the others. Under these conditions, one of the equations is dropped before the system (14) is estimated. In addition, the cross-equation symmetry restrictions $\gamma_{ij} = \gamma_{ji}$, and homogeneity restrictions $\sum_j \gamma_{ij} = 0$, should be tested, and imposed if accepted. Additional tests can be used to check that the estimated cost function is concave in wages.

¹⁷ Multiplying the *ex post* rental price of each capital asset in an industry times the stock of that asset, and summing over all types of capital, equals observed payments to capital (i.e. value of shipments less payments to labor and materials). The share of computers in the capital stock is

measured by taking the computer stock times its rental price, divided by the sum over all assets of each asset stock times its rental price. The *ex ante* rental prices are constructed by omitting capital gains on each asset, and using a “safe” rate of return. See Feenstra and Hanson (1999) for further details on these rental prices and the computer and other high-tech shares.

¹⁸ Autor, Katz, and Krueger (1998) obtain smaller estimates for the impact of outsourcing on the nonproduction wage share. This appears to be due to the fact that they also include as regressors industry measures of imports and exports, which are highly correlated with imported intermediate inputs.

¹⁹ This finding may be driven by the fact that through the 1980’s U.S. MNEs continued to concentrate their foreign operations in OECD countries, where their production activities are similar to those in U.S. manufacturing plants. During the 1990’s, this pattern began to change, as FDI by U.S. MNEs shifted towards emerging economies. Future foreign production transfer by U.S. MNEs may then be associated with skill upgrading in the United States. See Blomstrom, Fors, and Lipsey (1997) and Lipsey (1999).

²⁰ See also Brainard and Riker (1997) and Riker and Brainard (1997). Lipsey (1994) and Feliciano and Lipsey (1999) discusses employment and compensation practices of foreign MNEs in the United States.

²¹ Head and Ries also find that when Japanese firms move production to low-income countries they raise input purchases from these countries, but when they move production to high-income countries they do not. One interpretation of these results is that when Japanese firms move production to low-wage countries they mainly outsource low-skill tasks, which raises skill intensity at home, but when they move production to high-wage countries they tend to replicate

production activities done at home, which may lower employment in Japan but does not change skill intensity.

²² Their estimation equation is based on Feenstra and Hanson (1996), and differs somewhat from that in (15). In this specification, the level of foreign assembly activities is treated as an endogenous variable.

²³ We convert (18) to a discrete-time formula by replacing the instantaneous changes by the change in log prices between two years. In that case, we need to be specific about the year in which the factor shares are measured. A preferred method is to use the *arithmetic average* of the factor cost-shares in the two years, and this formulation is called the Tornqvist index of productivity. Caves, Christensen and Diewert (1982a,b) show that the Tornqvist index is a valid measure of Hicks-neutral or factor-biased productivity change – i.e., it is valid even when the shift parameters in the translog production or cost function have a non-neutral impact on factor demands.

²⁴ Feenstra and Hanson (1999) verify that the structural variables affect both TFP and value-added prices directly. Were the structural variables to affect just TFP, they would affect value-added prices indirectly, through the pass-through of productivity changes to product-price changes. Feenstra and Hanson find that the structural variables affect value-added prices over and above their impact on productivity.

²⁵ Without loss of generality, we also impose the symmetry restrictions $\gamma_{mn} = \gamma_{nm}$ and $\delta_{mk} = \delta_{km}$.

²⁶ Because the shares sum to unity, one product-share and one factor-share equation must be dropped.

Country	1974	1984	1993
All Manufacturing Industries			
Canada	15.9	14.4	20.2
Japan	8.2	7.3	4.1
United Kingdom	13.4	19.0	21.6
United States	4.1	6.2	8.2
Chemical and Allied Products			
Canada	9.0	8.8	15.1
Japan	5.2	4.8	2.6
United Kingdom	13.1	20.6	22.5
United States	3.0	4.5	6.3
Industrial Machinery (Non-electrical)			
Canada	17.7	21.9	26.6
Japan	2.1	1.9	1.8
United Kingdom	16.1	24.9	31.3
United States	4.1	7.2	11.0
Electrical Equipment and Machinery			
Canada	13.2	17.1	30.9
Japan	3.1	3.4	2.9
United Kingdom	14.9	23.6	34.6
United States	4.5	6.7	11.6
Transportation Equipment			
Canada	29.1	37.0	49.7
Japan	1.8	2.4	2.8
United Kingdom	14.3	25.0	32.2
United States	6.4	10.7	15.7

Note: US estimates are for 1975, 1985, and 1995.

Source: Campa and Goldberg (1997, Tables 1,3,5,7).

Table 1. Share of Imported to Total Intermediate Inputs (Percent)

	Domestic Price	Import Prices
<u>United States (1980-89)</u>		
All manufacturing industries		
Nonproduction labor weights	33.1	26.0
Production labor weights	32.3	28.1
<u>Japan (1980-90)</u>		
All manufacturing industries		
Nonproduction labor weights	-5.60	-18.23
Production labor weights	-3.90	-17.29
• Without Office Machines		
Nonproduction labor weights	-7.09	-18.69
Production labor weights	-4.72	-17.50
• Also without Petroleum Products		
Nonproduction labor weights	-6.98	-18.45
Production labor weights	-4.66	-17.39
<u>Germany (1980-90)</u>		
All manufacturing industries		
Non-manual labor weights	23.98	15.24
Manual labor weights	26.03	17.07
• Without Office Machines		
Non-manual labor weights	24.79	15.38
Manual labor weights	26.21	17.11
• Also without Petroleum Products		
Non-manual labor weights	24.97	15.70
Manual labor weights	26.28	17.24

Note: The averages shown weight each industry's price change by that industry's share of total manufacturing employment or nonproduction and non-manual workers, or production and manual workers. Industries are defined at the 3-digit standard industrial classification (SIC) level for the US, and generally correspond to the 2-digit level for Japan and Germany.

Sources: Lawrence and Slaughter (1993, Tables 3 and 4) and Lawrence (1994, Table 4).

Table 2. Employment Weighted Percentage Changes in Domestic and Import Prices

A. Industry Level Decomposition (percent)

<u>Year</u>	<u>Change in Employment</u>		<u>Change in Wages</u>	
	<i>Between</i>	<i>Within</i>	<i>Between</i>	<i>Within</i>
1973-79	0.121	0.199	0.119	0.212
Total	0.320		0.381	
1979-1987	0.184	0.362	0.309	0.410
Total	0.546		0.719	

B. Plant Level Decomposition (percent)

<u>Year</u>	<u>Change in Employment</u>		<u>Change in Wages</u>	
	<i>Between</i>	<i>Within</i>	<i>Between</i>	<i>Within</i>
1973-79	0.101	0.170	0.140	0.134
Total	0.271		0.274	
1979-1987	0.177	0.215	0.315	0.221
Total	0.392		0.536	

Note:

Numbers are percentage changes between years. *Between* numbers represent shifts across 4-digit SIC industries in part A, and shifts across plants in part B. *Within* numbers represent changes within industries in part A, and within plants in part B. All calculations have been annualized.

Sources:

Part A from Berman, Bound and Griliches (1993), and part B from Bernard and Jensen (1997).

Table 3. Employment and Wages of Non-Production Workers, 1973-79 and 1979-87

	(1) Mean	(2) Regression	(3) Regression	(4) Regression	(5) Contri- bution
$\Delta \ln(K/Y)$	0.706	0.047 (0.011)	0.044 (0.011)	0.040 (0.009)	7.3-8.5%
$\Delta \ln(Y)$	1.541	0.020 (0.006)	0.017 (0.006)	0.010 (0.006)	4.0-7.8%
Outsourcing	0.223	0.197 (0.096)	0.221 (0.100)	0.135 (0.088)	14.6-24.0%
<u>Computer and other high-tech capital measured with <i>ex post</i> rental prices:</u>					
Computer share	0.251	0.195 (0.091)			12.6%
Other High-tech share	0.144	-0.065 (0.137)			--
<u>Computer and other high-tech capital measured with <i>ex ante</i> rental prices:</u>					
Computer share	0.070		0.431 (0.167)		7.8%
Other High-tech Share	0.166		0.005 (0.071)		0.2%
<u>Computers measured as share of investment:</u>					
Computer share	6.561			0.018 (0.007)	30.5%
High-tech share (<i>ex post</i> rental prices)	0.395			0.032 (0.052)	3.3%
Constant		0.203 (0.043)	0.206 (0.040)	0.157 (0.045)	40.4-53.1%
R ²		0.156	0.159	0.189	
N		447	447	447	

Note: The mean of the dependent variable equals 0.389. Standard errors (in parentheses) are robust to heteroskedasticity and correlation in the errors within two-digit industries. The first column shows mean values of the dependent and independent variables for 1979-1990. All regressions and means are computed over 447 four-digit standard industrial classification (SIC) industries and are weighted by the average industry share of the manufacturing wage bill. $\Delta \ln(K/Y)$ is the average annual change in the log capital-shipments ratio and $\Delta \ln(Y)$ is the average annual change in log real shipments. The outsourcing variables and the computer and high-technology shares are in annual changes.

Table 4: Dependent Variable - Change in Nonproduction Wage Share, 1979-1990

	(1)	(2)	(3)	(4)	(5)
Effective TFP				-1.00 (0.007)	-1.00 (0.001)
TFP	-0.96 (0.07)	-0.75 (0.08)			
Production cost-share	3.06 (1.22)	2.43 (1.16)	3.61 (1.89)	4.68 (0.02)	4.70 (0.01)
Nonproduction cost-share	2.30 (1.43)	4.09 (1.72)	6.20 (4.04)	5.48 (0.02)	5.44 (0.03)
Capital cost-share	7.89 (0.78)	8.06 (0.94)	9.54 (2.19)	3.95 (0.01)	3.97 (0.02)
Materials cost-share times change in materials price	1.00*	1.00*	1.22 (0.25)	1.00*	1.00 (0.002)
Energy cost-share times change in energy price	1.00*	1.00*	-0.93 (0.92)	1.00*	1.00 (0.006)
constant	-0.71 (0.30)	-0.83 (0.29)	-1.93 (0.92)		0.01 (0.005)
R ²	0.896	0.806	0.429	0.999	0.999
N	447	446	446	447	447

Note: Standard errors are in parentheses. All regressions omit three industries with missing data on materials purchases or prices (SIC 2067, 2794, 3483) and are weighted by the industry share of total manufacturing shipments, averaged over the first and last period.

In columns (1)-(3) and (5), the dependent variable is the log change in the gross industry price, and the factor cost shares sum to one across all factors. The materials cost share is multiplied by the log change in the materials price; the energy cost share is treated similarly. In column (4), the dependent variable is the log change in the industry value-added price and factor cost shares sum to one across primary factors. Column (1) uses primal TFP as a regressor; column (2) drops the computer industry (SIC 3573) from the sample; column (3) also drops TFP as a regressor; and column (5) uses effective TFP as a regressor, where effective TFP equals primal TFP minus the change in factor-price differentials.

* These coefficients are constrained at unity.

Source: Feenstra and Hanson (1999).

Table 5: Dependent Variable - Log Change in Industry Price, 1979-1990