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The Diffusion of Mexican Immigrants during the 1990s
Explanations and Impacts

David Card and Ethan G. Lewis

During the 1990s the number of Mexican immigrants living in the United States rose by nearly five million people. This rapid growth is illustrated by the solid line in figure 6.1, which shows the number of working-age Mexican immigrants recorded in the 2000 Census by year of arrival in the United States.[1] At the time of the census, Mexican immigrants represented 4.1 percent of the working-age population, nearly double their proportion in 1990. The surge in arrivals from Mexico was accompanied by a remarkable shift in their residence patterns. In previous decades, nearly 80 percent of Mexican immigrants settled in either California or Texas. Over the 1990s, however, this fraction fell rapidly. As shown by the dotted line in figure 6.1, less than one-half of the most recent Mexican immigrants were living in California or Texas in 2000. Many cities that had very few Mexican immigrants in 1990—including Atlanta, Raleigh-Durham, Portland, and Seattle—gained significant Mexican populations. The inflow of Mexican immigrants to Southeastern cities is particularly significant because of the potential impact on the labor market prospects of less-skilled African Americans.

In this paper we explore potential explanations for the widening geo-

1. The jagged nature of the line reflects the tendency of census respondents to report that they arrived five, ten, fifteen, . . . years ago.
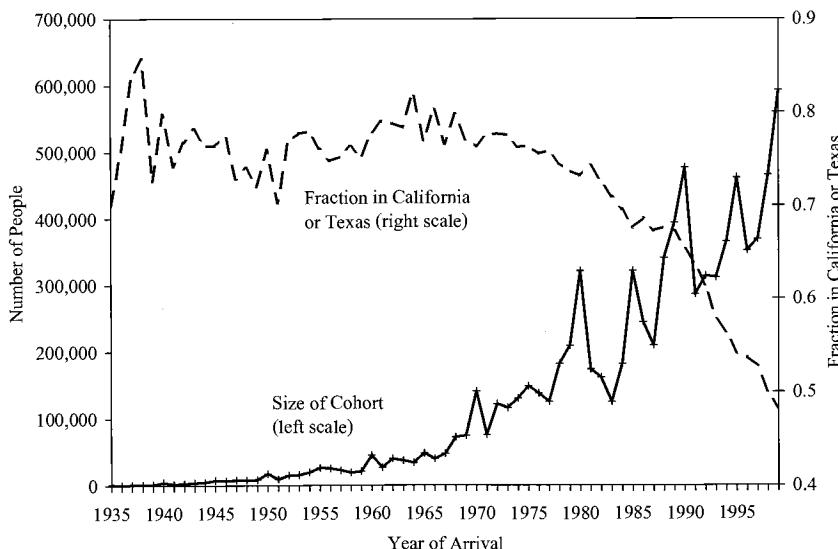
Fig. 6.1   Number and location of Mexican immigrants, by arrival year

graphic distribution of Mexican immigrants and examine the effects of Mexican immigration on local labor markets across the country. We begin with a descriptive overview of the location choices and other characteristics of recent Mexican immigrants. Post-1990 Mexican immigrants have about the same education and English-speaking ability as those who arrived in earlier decades. They differ mainly in their destinations: those who arrived in the 1990s were less likely to move to Los Angeles (the traditional destination of about one-third of all Mexican immigrants) and more likely to move to cities in the Southeast, Northwest, and Mountain states. The geographic shift was associated with some change in industry concentration, with fewer of the recent arrivals working in agriculture and more in construction (for men) and retail trade (for women).

We then go on to a more formal analysis of the role of supply-push and demand-pull factors in explaining the diffusion of Mexican immigrants across U.S. cities in the 1990s. Supplies of potential immigrants were rising over the decade, driven by population growth, falling real wages, and persistently weak economic conditions in Mexico.[2] Historically, new immigrants tend to follow earlier immigrants from the same country. Thus, we use information on the fraction of Mexican immigrants in a city in 1980 and 1990 as predictors of the supply-push component of immigrant flows.

2. Real wages in Mexico were about 20 percent lower in 2000 than in 1990. See Organization for Economic Cooperation and Development (OECD; 2000, 32).

On the demand side, we use predicted county-level employment growth over the 1990s, extrapolated from trends in the 1980s, as a measure of exogenous employment demand growth. Both factors are significant predictors of Mexican immigrant inflows, with supply-push factors explaining 75 percent of the intercity variation in inflow rates over the 1990s and demand-pull factors explaining another 10 percent.[3] By comparison, the relative wages and employment rates of Mexican immigrants in a city in 1990 are uncorrelated with subsequent inflows.

The remainder of the paper is focused on understanding how inflows of Mexican immigrants have affected local labor market conditions. We begin by showing that higher inflows of recent Mexican immigrants are associated with increases in the relative supply of less-educated labor in the local economy. Offsetting movements of previous immigrants and natives with low levels of education appear to be relatively small. We then examine the role of changing industry structure in explaining the absorption of relatively unskilled population inflows. The Hecksher-Olin (HO) model of trade suggests that shifts in the relative supply of unskilled labor can be absorbed by the expansion of low-skill-intensive industries, with little or no change in relative wages of unskilled workers. We develop a simple decomposition that allows us to characterize the fraction of the excess supply of dropout labor in a local market that has been absorbed by HO-style industry shifts. Our analysis suggests that between-industry shifts account for only a small fraction of the overall absorption of the extra dropout labor created by Mexican inflows.

In view of this finding, we turn to the impact of Mexican immigration on the relative wage structure. We construct estimates of the wage gap in each city between native men with exactly twelve years of schooling and those who did not complete high school and relate this gap to the relative supply of dropouts in the local market. Consistent with most of the existing literature (see, e.g., the review in Card 2005) we find that increases in the relative supply of dropouts induced by Mexican immigration inflows have small effects on relative wages of less-educated natives. The absence of a discernable effect on relative wages is especially puzzling given that most of the absorption of the excess supply of dropout labor created by Mexican immigrant inflows arises within narrowly defined (three-digit) industries. Evidently, the adjustments needed to accommodate differences in the relative supply of dropout labor in different markets occur without the intervening mechanism of relative wage changes. The data do not allow us to tell whether this is because high school dropouts and high school graduates are highly substitutable in production or as a result of other adjustment processes, such as endogenous technical change.

3. The two components are almost orthogonal so their contributions add up.

6.1   An Overview of Mexican Immigration in the 1990s

6.1.1   Census Data

Our empirical analysis is based on public use data from the 1980, 1990, and 2000 Censuses. The primary advantages of these data files are sample size and geographic coverage. For example, the 1980 Census includes 109,628 Mexican immigrants (72 percent of whom are between the ages of sixteen and sixty-five) and identifies more than 300 separate Metropolitan Statistical Areas (MSAs). A serious disadvantage is undercoverage of Mexican immigrants. Calculations by Borjas, Freeman, and Lang (1991) suggest that the 1980 Census missed approximately 40 percent of unauthorized Mexican immigrants, leading to a 25 percent undercount in the overall Mexican immigrant population.[4] Van Hook and Bean (1998) use a similar method to estimate a 30 percent undercount rate of unauthorized Mexicans in the 1990 Census and a 20 percent undercount of all Mexicans.[5] Analysts believe that the 2000 Census was substantially more successful in counting unauthorized immigrants (Citro, Cork, and Norwood 2004), with net undercount rates on the order of 10 percent (U.S. Department of Citizenship and Immigration Services 2003). This suggests an undercount rate for all Mexican immigrants of about 6–8 percent.[6] Based on these estimates, we believe that problems caused by the undercount of unauthorized Mexicans are likely to be relatively modest in our 2000 data, but more of an issue in interpreting the 1980 and 1990 data.

With these caveats in mind, we turn to table 6.1, which presents information on the characteristics of working-age Mexican immigrants in the 1980, 1990 and 2000 Censuses.[7] The demographic characteristics are fairly stable over time, though the average age of Mexican immigrants and their number of years in the United States are rising over time, reflecting the accumulating stock of previous migrants. There is also a modest upward trend in average education. Even in 2000, however, 70 percent report having less than a high school education, and more than one-half report low

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4. Estimates of the overall census undercount rates (based on sets of households that were identified and interviewed in two separate counts) are 1.2 percent for the 1980 Census, 1.6 percent for 1990, and 0.1 to 1.1 percent for 2000. Estimated undercount rates are higher for Hispanics (e.g., around 5 percent in the 1990 Census (Hogan and Robinson 1993), and 1–4 percent in the 2000 Census (Elliot and Little 2005). Estimates of undercount rates for the unauthorized population are based on comparisons of birth or death rates to population estimates.

5. Van Hook and Bean (1998) show the sensitivity of their estimates to various assumptions. The 30 percent undercount rate is based on relatively conservative assumptions. Other assumptions lead to lower undercount rates, on average.

6. Passel (2002) estimates that 80 percent of all Mexican immigrants who arrived in the 1990s were unauthorized.

7. We define *Mexican immigrants* as census respondents who report that they are either naturalized citizens or noncitizens and who report that their place of birth is Mexico.

Table 6.1   **Characteristics of Mexican immigrants in 1980, 1990, and 2000**

	1980	1990	2000
Percent female	46.5	44.2	43.7
Age distribution			
Percent under 30	47.2	45.5	39.9
Percent 31–50	40.9	44.2	49.2
Percent 51–65	11.9	10.3	10.9
Distribution of years in United States			
0–5 years	30.0	26.3	25.0
6–10 years	25.2	20.0	19.1
10 or more years	44.7	53.6	55.9
Education			
Percent <12 years schooling	76.7	74.6	70.2
Mean years of schooling	7.4	8.1	8.4
Percent low English ability	54.6	50.3	52.3
Geographic distribution			
Percent in California	58.0	58.4	44.9
Percent in Texas	22.2	21.2	19.5
Percent in MSA	92.8	91.3	90.1
Labor market outcomes			
Percent employed last year			
Men	85.9	85.7	83.9
Women	49.4	53.7	52.9
Mean hourly wage (1999$)			
Men	14.22	11.61	12.89
Women	11.06	9.68	11.07
Mean log wage gap relative to other workers ($\times$ 100)			
Men	–30.6	–42.6	–41.2
Women	–17.0	–29.5	–33.2
Percent of total population (age 16–65)	1.13	2.16	4.11
Sample size	83,628	174,364	373,909

Note: Based on tabulations of individuals age sixteen–sixty-five in 1980–2000 U.S. and Mexico Censuses.

or very low English ability.[8] The fraction of Mexican immigrants living in either California or Texas was stable between 1980 and 1990, but fell sharply in the 1990s. Roughly 90 percent of Mexican immigrants lived in a larger urban area (i.e., in a metropolitan area or consolidated metropolitan area) in 1980, and this rate has not changed much over the past two decades. Finally, the employment rates of Mexican immigrants have been relatively stable, whereas average real wages show a decline between 1980 and 1990 and a modest rebound by 2000. The hourly wage gap between

8. Based on observation at an English instruction class for immigrant parents, we suspect that Mexican immigrants tend to overreport their education. Many immigrants from rural areas attended ungraded schools with interruptions for work at home, so "years of school" may overstate actual years of full-time learning.

Mexican workers and all other workers expanded between 1980 and 1990 and was fairly stable between 1990 and 2000.

6.1.2   Intercohort Comparisons

Comparisons across the populations in different census years potentially mask differences between newly arriving and earlier cohorts of Mexicans. Figures 6.2–6.8 compare Mexican immigrants by years of residence in the United States in 1990 and 2000. A caveat in the interpretation of these figures is that many Mexican immigrants enter and leave the United States multiple times, leading to some ambiguity in the "arrival year" responses in the census. Moreover, some migrants enter and then leave permanently (Lubotsky 2000). To the extent that these factors are stable over time, however, comparisons by years since arrival in the different census years are informative.

Figure 6.2 plots the fractions of Mexican immigrants living in California and Texas by years in the country. In 1990, the probabilities of living in California or Texas were fairly similar for different arrival cohorts. In the 2000 data, however, recent arrivals are much less likely to live in California than earlier cohorts. This contrast suggests that the widening geographic diffusion of Mexican immigrants during the 1990s was driven by the locational choices of new immigrants—a conclusion that is reinforced by further analysis in the following.

Figures 6.3 and 6.4 compare the fractions of Mexican immigrants with less than a high school degree and with low English ability. Female immigrants from Mexico have about the same probability of below-high school
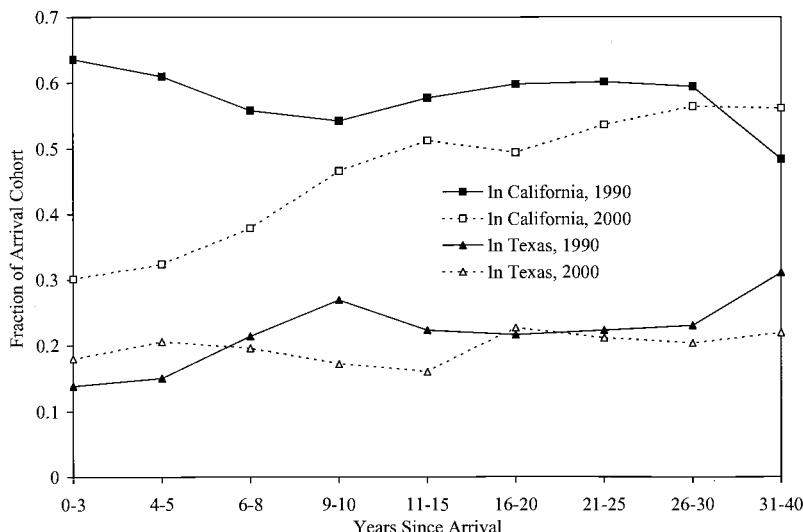

**Fig. 6.2   Location of Mexican immigrants, by years since arrival**
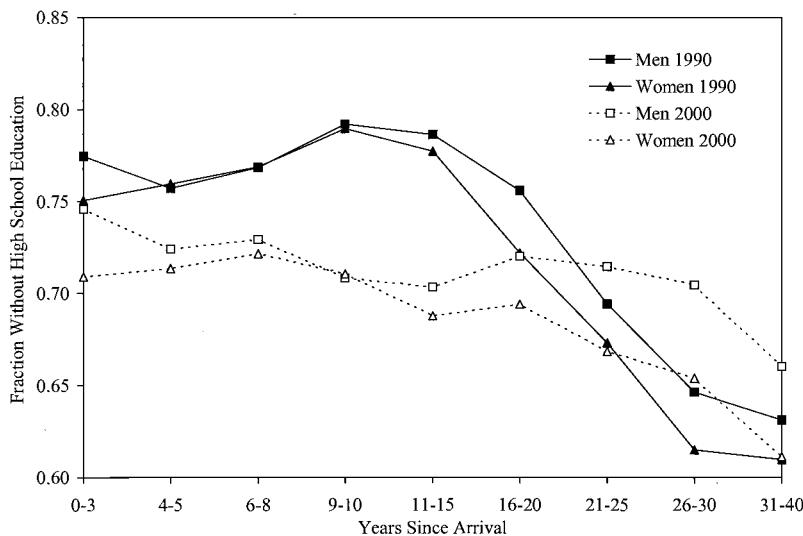
Fig. 6.3   Fraction of Mexican immigrants with less than high school education, by years since arrival
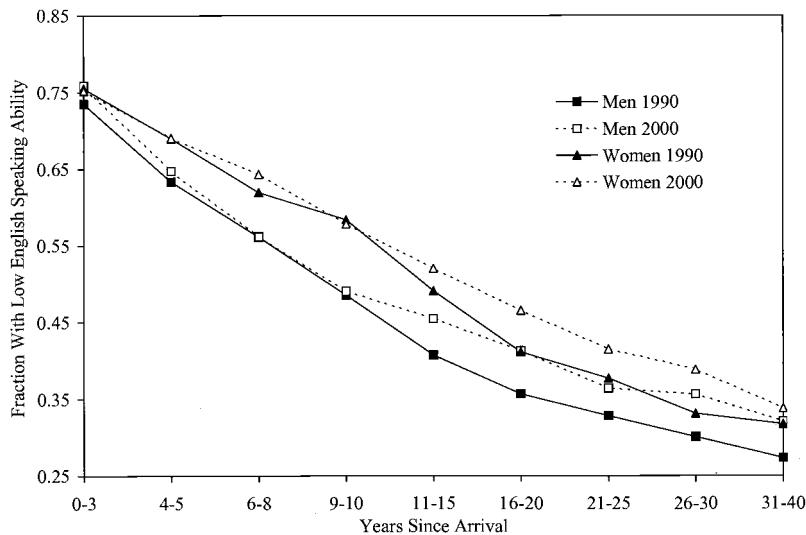
Fig. 6.4   Fraction of Mexican immigrants with low English, by years since arrival

education as males, but report lower English-speaking abilities. Recent arrivals of either gender in the 2000 Census have a slightly lower proba- bility of below-high school education than their counterparts in 1990, per- haps reflecting gains in education for younger cohorts in Mexico. The levels of low English ability, on the other hand, are very similar in 1990 and

2000.[9] Although we do not present them here, the marital status profiles for men and women are also remarkably similar in the two censuses.

Figure 6.5 shows mean log hourly wages (in 1999 dollars) by gender and time in the United States. There was a modest rise in real wages for more recent arrivals over the 1990s, but not much gain for longer-term residents. Overall, the wage profiles are quite similar in 1990 and 2000. We have also constructed profiles of employment probabilities (based on the likelihood of reporting positive weeks of work in the past year). For men, the 1990 and 2000 profiles are very close together, while for women there is a slightly lower employment rate in 2000 for those who have been in the United States for six–ten years and not much difference elsewhere.[10]

Finally, figures 6.6–6.8 show the fractions of Mexican workers employed in agriculture, construction, and retail trade. In 1990, the data in figure 6.6 show that recently arrived Mexicans of either gender were more likely to work in agriculture than earlier arrivals. (Of course, this could have been driven by the presence of many short-term migrant workers in agriculture in 1990.) By 2000, however, the profiles by time in the United States are much flatter. Looking across major industry groups, we found that the decline in agricultural employment among recent immigrants was offset by rises in the fraction of employment in construction (for men) and retail trade (for women). In 2000, nearly a quarter of recent male Mexican immigrants was working in construction (see figure 6.7), while about one-sixth of recent females were working in retail trade (figure 6.8). The rises in Mexican employment in these industries are striking because both sectors also employ relatively large fractions of low-skilled native workers, raising the obvious concern about labor market competition.

6.1.3   Distribution across Cities

As we have noted, one of the most important changes for Mexican immigrants between 1990 and 2000 was the move out of California. Further information on this phenomenon is provided in table 6.2, which shows the changing fractions of Mexican immigrants in the fifteen traditional destination cities that had the largest numbers of Mexicans in 1980.[11] In 1980, nearly one-third of all working-age Mexicans were living in Los Angeles.

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9. To the extent that the immigrants who are most likely to be undercounted in the census are recent arrivals with low education and language ability, there may be more reporting bias in the 1990 Census data than the 2000 data. This would tend to mask any actual gains in education or English ability that actually occurred over the 1990s.

10. As with education and language, there may be some correlation between wages and the probability of underreporting, especially for recent Mexican immigrants. Assuming this was a bigger problem in 1990, the observed mean wage trends for recent arrival groups may understate the actual growth that occurred.

11. Throughout this paper, we use as *cities* individual MSAs and the constituent PMSA's in consolidated metropolitan areas. Thus, we treat Los Angeles and Orange County California as separate *cities*.
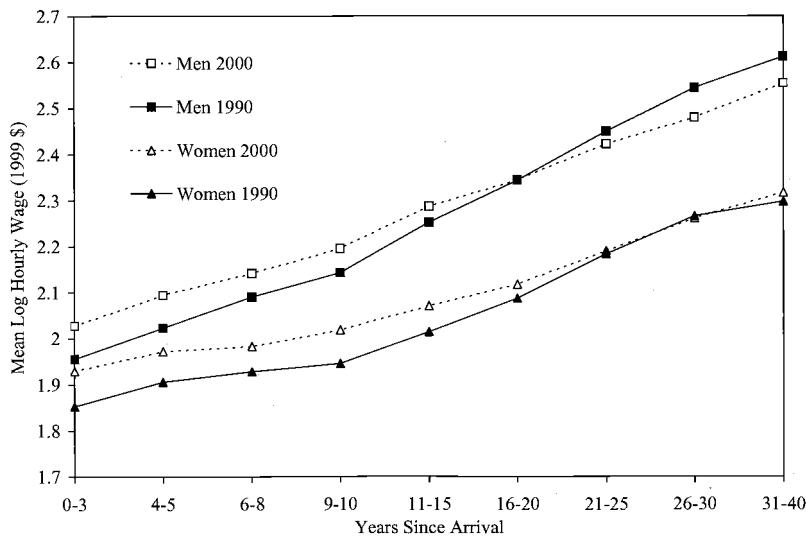
Fig. 6.5   Mean log hourly wages of Mexican immigrants, by years since arrival
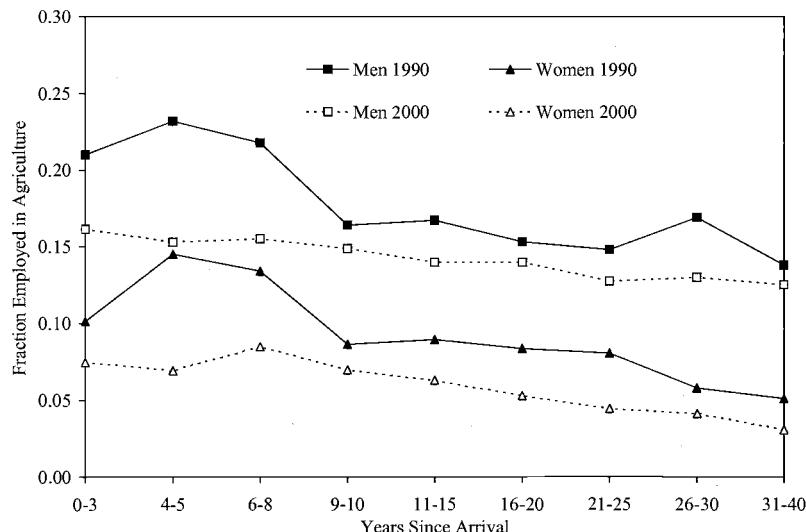
Fig. 6.6   Fraction of Mexican immigrants in agriculture, by years since arrival

Another 8 percent were living in Chicago, and roughly 4 percent were living in each of Houston, Orange County, San Diego, and El Paso. Over the 1980s, the shares in Los Angeles and Chicago fell slightly, but as of 1990 the top five cities still accounted for nearly one-half of all Mexican immigrants. Between 1990 and 2000, however, the share of Mexican immigrants
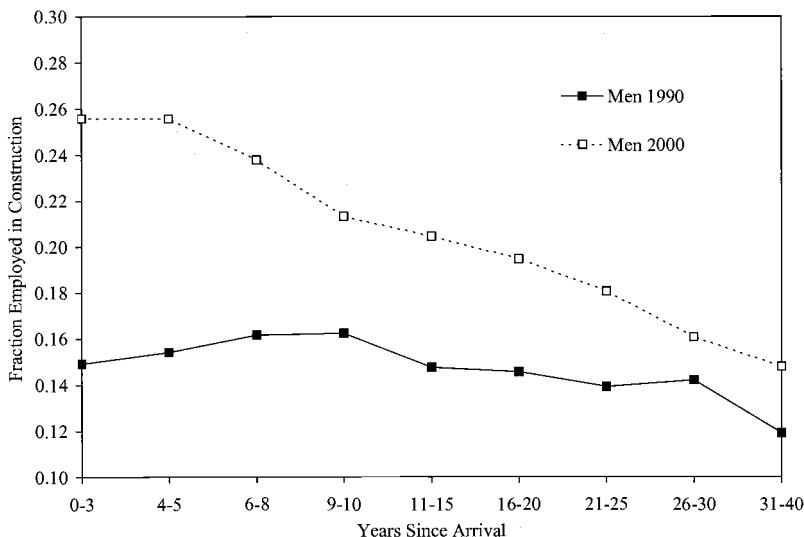
Fig. 6.7   Fraction of Mexican immigrants in construction, by years since arrival
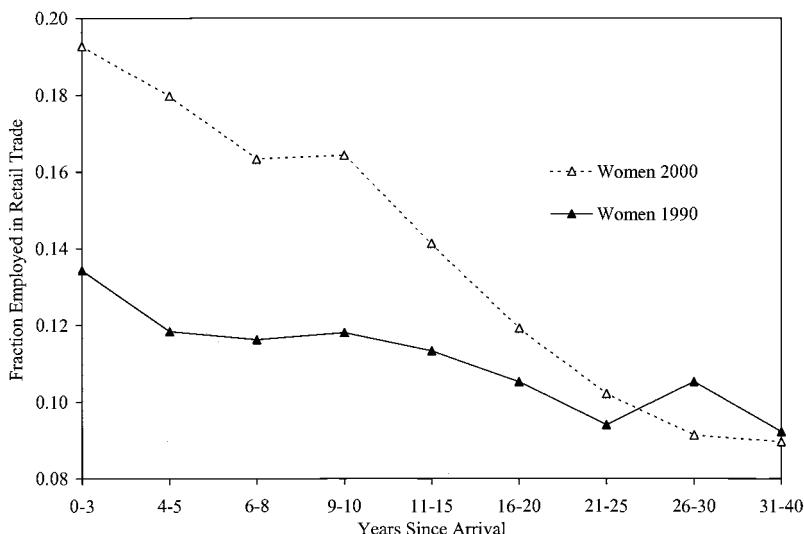
Fig. 6.8   Fraction of Mexican immigrants in retail trade, by years since arrival

living in Los Angeles dropped by 10 percentage points, accounting for most of the fall in the total California share noted in figure 6.1 and table 6.1. The total share in Texas fell by much less, although this stability masks a sizeable (2.8 percentage point) loss in shares for San Antonio and the smaller border cities (El Paso, McAllen, and Brownsville) coupled with gains for Houston and Dallas.

Table 6.2   **Geographic concentration of Mexican immigrants**

	1980	1990	2000
Percent of Mexican immigrants (age 16–65) living in:			
Los Angeles	31.7	27.9	17.4
Chicago	7.9	5.4	5.5
Houston	4.4	4.1	4.4
Orange County, CA	4.1	6.0	4.7
San Diego	3.9	4.1	3.1
El Paso	3.9	2.7	1.6
San Francisco/Oakland	2.5	2.3	2.4
Dallas/Fort Worth	2.3	3.3	4.7
McAllen	2.1	1.7	1.5
San Antonio	2.0	1.5	1.1
San Jose	1.7	1.7	1.5
Brownsville	1.6	1.9	0.8
Ventura County, CA	1.6	1.4	1.1
Fresno	1.4	1.6	1.6
Riverside/San Bernardino, CA	1.3	4.1	4.1
Share of Top 5	51.9	47.5	35.1
Share of Top 15	72.3	69.7	55.5

Note: Based on tabulations of 1980–2000 U.S. and Mexico Censuses.

Where did the rapidly growing population of Mexican immigrants settle in the 1990s? To answer this question, we calculated the increase in the number of Mexican immigrants in each MSA between 1990 and 2000 and then tabulated the cities by their shares of the total increase in Mexican immigrants. The results for the top forty cities, which together account for about 80 percent of the overall growth in the Mexican population, are presented in table 6.3.

The first three columns of the table show the total working-age population of each city in 1990, the number of Mexican immigrants in 1990, and the fraction of Mexican immigrants in the local working-age population. The remaining five columns present information on the changes in each city between 1990 and 2000, including the total population growth rate (for sixteen- to sixty-five-year-olds), the growth rate of the Mexican immigrant population, the absolute increase in the total number of Mexican immigrants living in the city, the fraction of the national increase in the Mexican population "absorbed" in the city, and, finally, the number of post-1990 immigrants living in the city in 2000.

Although Los Angeles's share of Mexican immigrants was falling over the 1990s, the first row of table 6.3 shows that the city still absorbed the largest number of Mexicans (over 300,000). In fact, the Mexican population of Los Angeles grew by 34 percent between 1990 and 2000. Because the total population of Mexican working-age immigrants grew by 114 percent over the decade, however, Los Angeles would have had to absorb

Table 6.3   Growth in overall and Mexican immigrant populations, 1990 to 2000

	Working age population in 1990				Changes from 1990 to 2000			
	Total	Mexican immigrants	Percent Mexican immigrants	Adult population growth (%)	Mexican immigrant growth (%)	Growth in no. Mexican immigrants	Cumulative percent of total rise	No. of post-1990 Mexican immigrants
1 Los Angeles	5,785,200	973,120	16.8	5.8	33.5	326,260	9.5	413,140
2 Chicago	4,170,420	186,800	4.5	9.8	119.8	223,800	15.9	182,560
3 Phoenix	1,268,280	52,400	4.1	62.1	363.9	190,700	21.5	135,640
4 Dallas	1,693,060	85,320	5.0	36.6	215.6	183,960	26.8	153,240
5 Houston	1,908,400	137,320	7.2	26.2	131.3	180,300	32.0	147,220
6 Riverside/San Bernardino, CA	1,444,480	142,620	9.9	27.8	112.6	160,560	36.6	89,860
7 Orange County	1,687,500	208,000	12.3	13.1	69.7	144,900	40.8	139,200
8 Las Vegas	470,200	15,240	3.2	113.4	615.8	93,840	43.6	52,360
9 San Diego	1,603,060	144,440	9.0	11.1	63.0	90,980	46.2	74,960
10 Atlanta	1,757,700	7,320	0.4	41.7	1,076.5	78,800	48.5	66,180
11 New York City	4,520,040	28,140	0.6	13.8	270.2	76,020	50.7	67,040
12 Denver	1,018,560	12,620	1.2	25.5	585.4	73,880	52.8	54,260
13 Oakland	1,379,320	46,500	3.4	13.7	152.2	70,780	54.9	52,780
14 Fresno	396,560	54,320	13.7	41.4	113.7	61,780	56.7	45,260
15 Fort Worth	920,880	30,240	3.3	10.6	178.3	53,920	58.2	41,700
16 Austin, TX	518,400	16,060	3.1	60.9	322.0	51,720	59.7	36,900
17 San Jose	1,048,420	59,660	5.7	8.0	86.0	51,280	61.2	49,380
18 McAllen, TX	202,860	60,960	30.1	45.5	84.1	51,240	62.7	35,900
19 Tulare County, CA	171,560	31,800	18.5	67.4	154.3	49,080	64.1	28,800

20	Monterey County, CA	224,860	36,000	16.0	43.2	131.5	47,340	65.5	31,420
21	Bakersfield	318,120	32,480	10.2	22.4	124.8	40,540	66.6	25,060
22	Portland	885,080	9,320	1.1	35.1	424.3	39,540	67.8	30,820
23	Ventura County, CA	437,260	48,200	11.0	9.3	76.8	37,020	68.9	28,920
24	San Francisco	1,030,900	32,720	3.2	12.6	91.0	29,760	69.7	28,740
25	Raleigh-Durham	469,180	880	0.2	78.0	3,156.8	27,780	70.5	23,360
26	San Antonio	841,060	51,400	6.1	4.2	53.4	27,420	71.3	26,240
27	El Paso	351,640	93,900	26.7	5.2	28.3	26,580	72.1	32,140
28	Greensboro, NC	729,680	1,140	0.2	8.8	2,287.7	26,080	72.8	21,420
29	Salt Lake City	575,160	3,700	0.6	23.4	670.3	24,800	73.6	18,960
30	Sacramento	941,920	22,040	2.3	6.6	112.4	24,780	74.3	19,200
31	Santa Barbara	251,580	27,460	10.9	15.5	88.1	24,200	75.0	21,640
32	Tucson	399,780	23,880	6.0	28.7	100.8	24,060	75.7	16,900
33	Seattle	1,204,960	3,360	0.3	18.0	695.8	23,380	76.4	17,140
34	Washington, DC	2,610,900	7,860	0.3	25.1	271.3	21,320	77.0	18,260
35	Stockton, CA	298,380	26,380	8.8	19.5	79.9	21,080	77.6	18,800
36	Charlotte, NC	785,040	900	0.1	12.7	2,313.3	20,820	78.2	17,660
37	Yuma, AZ	59,120	12,920	21.9	59.5	158.5	20,480	78.8	10,600
38	Modesto, CA	231,440	21,460	9.3	22.0	91.4	19,620	79.3	13,420
39	Santa Rosa	224,040	8,840	4.0	40.2	217.7	19,240	79.9	13,700
40	Vallejo, CA	302,080	11,560	3.8	10.2	159.2	18,400	80.4	13,500

nearly a million Mexicans to maintain its share. In contrast to Los Angeles, Chicago's Mexican immigrant population grew at about the national average rate, implying a near doubling of the Mexican immigrant share over the 1990s. Dallas and Houston had even faster growth rates in their Mexican populations, together absorbing nearly 10 percent of the national rise. Phoenix and Las Vegas—two very rapidly growing cities—also experienced rapid growth in their Mexican immigrant populations.

More surprising than these figures are the large numbers of Mexican immigrants absorbed in Atlanta, New York, and Denver (the cities ranked numbers 10–12 in table 6.3). All three cities are far from the Mexican border and had very low Mexican population densities in 1990, yet together these cities absorbed over 9 percent of the total increase in the Mexican immigrant population. Looking further down the table, Portland Oregon (22), Salt Lake City (29), Seattle (32), Washington, D.C. (34) and three cities in North Carolina—Raleigh-Durham (25), Greensboro (28), and Charlotte (36)—also stand out as cities with historically small Mexican immigrant populations that experienced very rapid inflows over the 1990s. Together these ten cities accounted for 412,000 of the rise in the adult Mexican population between 1990 and 2000, or 12 percent of the national total.

A key feature of table 6.3 is the high correlation across cities between the growth in the total number of working age Mexican immigrants (column [6]) and the number of post-1990 Mexican immigrants present in 2000 (column [8]). This correlation has two implications. On one hand, it suggests that the arrival of new Mexican immigrants had little displacement effect on previous Mexican immigrants. On the other, it also implies that most of the growth in the number of Mexicans in new destination cities was attributable to the arrival of recent immigrants. These impressions are confirmed by the patterns in figure 6.9, which plots the change in the total number of adult Mexican immigrants living in each city between 1990 and 2000 (as a percent of the city's population in 1990) against the inflow rate of new Mexican immigrants, which we define as the number of post-1990 Mexican immigrants in the city in 2000 divided by the city population in 1990. The points for all but two cities lie on or above the 45-degree line, implying that in most cities new Mexican inflows led to equivalent or larger increases in the total Mexican population.[12] Only in Los Angeles and El Paso is there any evidence of displacement of older Mexican immigrants by new arrivals. In the labeled cities above the 45-degree line, net inflows of older immigrants complemented the inflows of post-1990 arrivals, amplifying the impact on local population growth.

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12. The same conclusion emerges when we plot the data for the 150 largest cities in the United States. Over this broader set, only three cities have notably smaller growth in the total Mexican population than in new Mexican inflows: Los Angeles; El Paso; and Laredo, Texas.
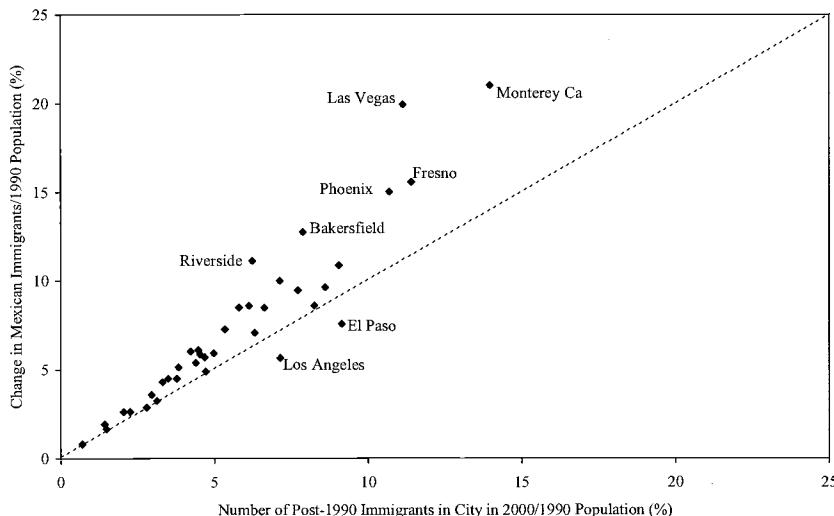
Fig. 6.9   Growth in Mexican immigrant population: Recent arrivals and total change

6.2   Modeling the Diffusion of Recent Mexican Immigrants

In light of this descriptive evidence, we turn to the task of modeling the flows of recent Mexican immigrants to different cities between 1990 and 2000. Our dependent variable is the inflow rate of new Mexican immigrants, defined as the number of post-1990 working age Mexican immigrants observed in a city in the 2000 Census, divided by the working age population of the city in 1990. Following the traditional taxonomy, we develop a framework for measuring the contribution of supply push and demand pull to total immigrant inflows. We measure demand-pull factors by total employment growth in the MSA between 1990 to 2000, derived from County Business Patterns (CBP) data.[13] There is a potential endogeneity problem with this variable as immigrant arrivals may stimulate employment growth. Exploiting the persistence in city-specific employment trends, however, we use employment levels from 1982 to 1990 as instruments for the 1990–2000 employment growth rate. Thus, our demand-pull measure is the predicted component of overall employment growth in the city, based on employment trends in the preceding decade.

On the supply side, numerous studies have shown that new immigrants tend to go to cities where earlier waves of immigrants from the same source

13. Except in New England, MSAs consist of complete counties, so MSA employment is the sum of employment in the constituent counties. For consistency, we use fixed 2000 MSA-county definitions.

country have settled (e.g., Bartel 1989; Card 2001). Thus, we use the density of Mexican immigrants in a city in 1980 and 1990 as proxies for the magnitude of supply-push immigration flows from Mexico over the 1990–2000 period.

Estimation results from a series of alternative specifications of the model are presented in table 6.4. The models are estimated on a sample of 142 larger MSAs that can be consistently defined on a county basis in the 1980, 1990, and 2000 Censuses.[14] The first column of the table reports a specification that includes only the lagged Mexican immigrant density variables. These supply-push proxies are highly significant and together explain 78 percent of the variation across cities in the recent Mexican immigrant inflow rate. The second column reports a model that includes only the employment growth variable. This is also a significant determinant of new immigrant inflows, explaining about 10 percent of the intercity variation. A parallel model estimated by instrumental variables is presented in column ([5]; using log employment levels in 1984–1990 as instruments). Interestingly, the point estimate of the effect of employment growth is slightly *larger* in the IV model, contrary to what might have been expected under the assumption that the ordinary least squares (OLS) estimate is upward biased by the presence of unobserved factors that contribute to both overall employment growth and Mexican inflows.[15] Finally, the models in columns (3) and (6) include both the lagged density and employment growth variables. Together the demand-pull and supply-push variables explain 86 percent of the intercity variation in new Mexican immigrant inflows. Again, the point estimates of the models are not much different between the OLS and IV specifications.[16]

Given the large fraction of Mexican immigrants who traditionally migrated to Los Angeles and the sharp decline in this fraction over the 1990s, an interesting challenge for our model is to predict the changing flows to Los Angeles. To address this challenge, we reestimated the model in column (3), adding a dummy for the Los Angeles observation. The estimated Los Angeles dummy is –0.025, with a standard error of 0.013, while the point estimates of the other coefficients are virtually the same as those reported in column (3). Thus, our baseline model overpredicts the inflow rate

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14. Copies of the computer programs that process the 1980, 1990, and 2000 Census data and construct the city-level variables are available on request.

15. The OLS estimate is probably downward biased by measurement errors in the CBP data. The CBP data are based on counts of people paying Social Security contributions and missing uncovered employment. Errors can also arise because of changes in the boundaries of MSAs between 1990 and 2000 and because of the fact that we measure population changes by place of residence, whereas CBP measures employment changes by place of work. It appears that the downward bias dominates any upward endogeneity bias.

16. The OLS estimate of the demand coefficient in column (3) is 0.0748, with a standard error of 0.008. The corresponding IV estimate in column (6) is 0.0675, with a standard error of 0.012.

Table 6.4   **Regression models for growth in recent Mexican immigrant population**

	Estimated by OLS				Estimated by IV		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Mexican population share, 1990	1.33		1.34	1.34		1.34	1.34
	(0.12)		(0.09)	(0.09)		(0.09)	(0.09)
Mexican population share, 1980	–1.16		–1.18	–1.18		–1.18	–1.18
	(0.18)		(0.14)	(0.14)		(0.14)	(0.14)
Employment growth, 1990–2000		0.09	0.07	0.08	0.12	0.07	0.07
		(0.02)	(0.01)	(0.01)	(0.03)	(0.01)	(0.01)
Mean log wage of Mexican men in 1990 ($\times 100$)				0.47			0.35
				(0.43)			(0.45)
Relative employment rate of Mexican men in 1990 ($\times 100$)				–0.12			–0.06
				(1.10)			(1.16)
R^2	0.78	0.11	0.86	0.86	0.10	0.86	0.86
First stage F-statistic (9 d.f.)					13.40	14.81	13.06

Notes: All models estimated on sample of 142 larger cities with census data for 1980–2000 and matching employment data from county business patterns for 1982–2000. Dependent variable is number of recent (post-1990) adult Mexican immigrants in city in 2000, divided by population in 1990. Mean and standard deviation of dependent variable are 0.023 and 0.033, respectively. Instruments for employment growth 1990–2000 are log employment levels in 1982–1990. Mean log wage and relative employment rate for city in 1990 are regression adjusted for characteristics of Mexican male workers in the city.

of new Mexican immigrants to Los Angeles (predicted inflow rate = 0.096; actual = .071), though the magnitude of the prediction error is just on the margin of statistical significance. Moreover, the Los Angeles observation is not a large enough outlier to have any affect on the coefficient estimates. The model in column (3) predicts that Los Angeles would have attracted about 558,000 new Mexican immigrants over the 1990s, compared to the actual inflow of 413,000. By comparison, if Los Angeles had maintained its 1990 share of Mexican immigrants, it would have attracted 961,000 new Mexican immigrants (an inflow rate of 0.165).[17] Thus, the decline in the share of Mexican immigrants moving to Los Angeles in the 1990s is largely explained by a combination of slow employment growth in the city and the pattern of the coefficients on lagged immigrant shares, which indicate a tendency for all cities with a longer history of Mexican immigration to have slower growth in new arrivals in the 1990s. An interesting question that we leave unanswered is whether this pattern could have been predicted by observing settlement patterns over the 1980s or whether it is a "new" phenomenon.

Although the simple supply-push and demand-pull proxies used in the models in columns (3) and (6) explain much of the variation in new Mexi-

17. Los Angeles had 27.9 percent of all working-age Mexican immigrants in 1990. According to the 2000 Census, there were 3,445,000 working age Mexicans who arrived after 1990 in the United States in 2000.

can immigrant inflow rates, other factors may also affect the destination choices of potential migrants. An obvious consideration is the labor market success of earlier cohorts of Mexican immigrants in a particular city. We used 1990 Census data to estimate the average employment rate and mean log wage of Mexican male immigrants in each city in 1989 (adjusted for the characteristics of the Mexican workers in each city).[18] We then included these as additional explanatory variables in the models in columns (4) and (7) of table 6.4. The results suggest that new immigrants tend to go to cities where Mexicans earned higher wages in 1990, although the estimated effects small in magnitude and insignificantly different from 0.[19] The estimated employment effects are also very small in magnitude and insignificantly different from 0. Overall these variables add little to our basic specification.

The models in table 6.4 are estimated using unweighted OLS and IV methods. We have also estimated the same specifications using weighted OLS and IV, with the MSA population in 1990 as a weight. The estimated coefficients from the weighted models are similar to the estimates from the unweighted models and lead to very similar conclusions about the explanatory power of the supply-push and demand-pull variables. As in the unweighted models, the weighted IV estimates of the employment growth effect are very close to the weighted OLS estimates, giving no indication of an endogeneity problem.

We conclude that a simple model that includes demand-pull and supply-push factors provides a relatively good description of the destination choices of new Mexican immigrants over the 1990s. A model with just three parameters explains 86 percent of the observed intercity variation in new Mexican immigrant inflow rates. The model cannot fully explain the sharp downturn in the share of Mexican inflows to Los Angeles in the 1990s, but it predicts about 75 percent of the observed decline.

6.3   Impacts of Mexican Inflows

6.3.1   Effects on the Relative Supply of Low-Education Labor

Having documented the relatively large inflows of Mexican immigrants to many cities in the 1990s, we now turn to analyzing the effects of these inflows. A first question is whether inflows of Mexican immigrants lead to any shift in the skill mix of local populations. Many models of local labor

18. To estimate these adjusted outcomes, we fit models for log hourly wages and the event of working last year, which included education, age, years in the United States, an indicator for low English ability, and unrestricted city dummies. We then use the city dummies as measures of relative wages and employment probabilities.

19. For example, the 0.47 coefficient for wages in the model in column (4) implies that cities where Mexican men earned 10 percent higher wages in 1989 had an inflow rate 0.0005 points higher in the 1990s.

market equilibrium have a constant-returns-to-scale feature that implies that population inflows only affect wages and employment to the extent that they shift the relative supply of different skill groups.[20]

As a starting point, figure 6.10 plots the change in the fraction of dropouts in the population of each major MSA between 1990 and 2000 against the inflow rate of new Mexican immigrants to the city. If 70 percent of recent Mexican arrivals have less than a high school education and Mexican inflows are orthogonal to all other characteristics in a city, then one would expect the points in figure 6.10 to lie along a line with slope slightly below 0.7.[21] For reference, we have graphed a line with this slope in the figure. While there is considerable variation in the scatter of points, there is a strong positive relation between Mexican inflows and the change in the dropout share, with a slope that is a little flatter than the reference line.

Table 6.5 presents a series of regression models that examine more formally the link between Mexican immigrant inflows and the share of low-education workers in a city. The dependent variable for the models in the first two columns is the fraction of dropouts among adult residents of a city in 2000, while in columns (3)–(5) the dependent variable is the *change* in the share of dropouts between 1990 and 2000. Looking first at the simple model in column (1), each percentage point increase in the inflow rate of new Mexican immigrants over the 1990s is estimated to raise the fraction of dropouts by 1.29 percentage points. This estimate is too large to represent a causal effect of the Mexican inflow. The problem is that inflows tend to be larger in cities that had larger inflows of Mexican immigrants in the past. This is illustrated by the model in column (2), which also includes the Mexican inflow rate over the 1980s. The 0.69 coefficient on the 1980s inflows suggest that Mexican arrivals have a highly persistent impact on the fraction of dropouts present in the labor market in 2000. Controlling for these flows, the marginal impact of inflows in the 1990s is about 0.9.

A potentially better specification relates the *change* in the dropout share to the inflow rate of new Mexican immigrants (as in figure 6.10). As shown by the models in columns (3) and (4), in such a specification each percentage point increase in the inflow rate of new Mexican immigrants is estimated to raise the fraction of dropouts in a city by 0.5 points. This estimate

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20. Strictly speaking, such a feature requires perfectly elastic supplies of capital to different cities and no shortage of land within a city. Arguably both features are true for many MSAs, though not necessarily for high density MSAs like Los Angeles or New York.

21. Let D_t represent the number of working age dropouts in a city in 2000, let N_t represent total working age population, let D_{t-1} and N_{t-1} represent the same concepts for 1990, and let M_t represent the number of new Mexican immigrants present in 2000. If 70 percent of new Mexican immigrants are dropouts, then $(D_t - D_{t-1})/N_{t-1} = .7\ M_t/N_{t-1} + \Delta/N_{t-1}$, where Δ represents the net change in the number of dropouts from all other sources (net flows of natives or previous immigrants). If Δ/N_{t-1} is orthogonal to the *inflow rate* of new Mexican immigrants, then a regression of $(D_t - D_{t-1})/N_{t-1}$ on M_t/N_{t-1} will have a coefficient of 0.7. The y-axis in the graph is $D_t/N_t - D_{t-1}/N_{t-1} = (D_t - D_{t-1})/N_{t-1} - (D_t/N_t) \times (N_t - N_{t-1})/N_{t-1}$. The second term in this sum is negatively correlated with the inflow rate of new Mexicans, leading to the prediction of a slope under 0.7.
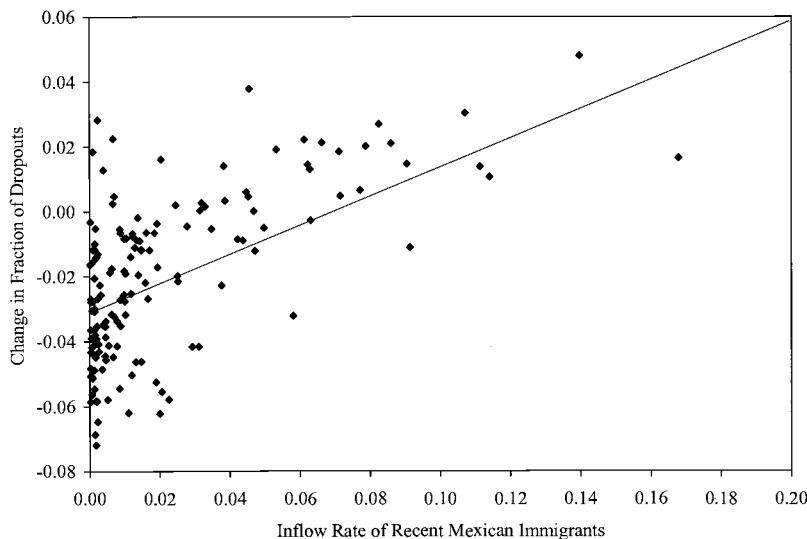
Fig. 6.10   Inflow rate of Mexican immigrants and change in fraction of dropouts

suggests that any offsetting migration of earlier immigrants or native dropouts induced by the inflow of new Mexican immigrants is relatively small. Interestingly, the inflow rate of immigrants in the 1980s has no effect on the change in dropout shares between 1990 and 2000, providing a simple specification check for the first-differenced model.

A concern with the models in columns (3) and (4) is that Mexican immigrants may be attracted to cities where there is an unusually high rate of growth in demand for less-educated labor. If that is the case and if less-educated natives (or less-educated immigrants from other countries) are attracted by the same demand factors, then the measured effect of Mexican inflows on the change in the dropout share may overstate their true net impact. Such a bias can be reduced or eliminated by using the supply-push variables (i.e., the historical fractions of Mexican immigrants in the city) as instruments for the inflow rate of new Mexican immigrants over the 1990s. We implement this procedure in the model in column (5). At the same time, we instrument employment growth in the city with the lagged employment variables used in table 6.4. The resulting coefficient estimates are not very different from the OLS estimates and provide no evidence that endogeneity of Mexican immigrant inflows leads to an overstatement of the effect of these flows on the relative fraction of dropout labor in a city. Overall, we conclude there is robust evidence that inflows of Mexican labor increase the share of dropouts in a city, with each percentage point increase in the inflow rate of recent immigrants leading to about a .5 percentage point higher dropout share in 2000.

Table 6.5   **Regression models for level or change in fraction of dropouts in local population**

| | Models for fraction of dropouts in 2000: Estimated by OLS | | Models for change in fraction of dropouts between 1990 and 2000: | | |
| | | | Estimated by OLS | | IV |
	(1)	(2)	(3)	(4)	(5)
Growth in "new" (post-1990) Mexican immigrants (1990–2000), divided by population in 1990	1.29 (0.11)	0.89 (0.19)	0.49 (0.04)	0.49 (0.10)	0.52 (0.05)
Growth in "new" (post-1980) Mexican immigrants (1980–1990), divided by population in 1980		0.69 (0.20)		0.01 (0.10)	
Employment growth, 1990–2000		–0.09 (0.03)		0.00 (0.01)	0.02 (0.01)
R^2	0.51	0.64	0.44	0.44	0.41

Notes: All models estimated on sample of 144 larger cities with census data for 1980–2000 and match-ing employment data from county business patterns for 1982–2000. Dependent variable is fraction of dropouts in adult population in city in 2000 (columns [1]–[2]) or the change in the fraction of dropouts in the adult population from 1990 to 2000 (columns [3]–[5]). Model in column (5) is estimated by in-strumental variables, using as instruments the fraction of Mexicans in the city in 1980 and 1990 and the log of employment in the MSA in 1982–1990.

6.3.2   Industry Structure and the Absorption of Mexican Labor

Because inflows of Mexican labor increase the pool of less-educated la-bor in a city, it is interesting to ask how these workers are absorbed by lo-cal employers. One possibility, suggested by the HO model of international trade, is that the industry structure in a city adapts to the relative supply conditions in the local labor market. Indeed, under certain conditions, changes in industry structure can fully accommodate differences in the rel-ative supply of different skill groups in a given city with no change in the relative wage structure. In this section, we use the decomposition method of Lewis (2003) to evaluate the role of HO-style adjustments in absorbing differences in the fraction of low-education workers in different cities.

The decomposition starts with an identity that expresses the overall frac-tion of dropouts employed in a given city, $s^d(c)$, as a weighted sum of the industry shares in the city, times the dropout intensity in each industry:

(1)
$$s^d(c) = \frac{1}{N(c)} \sum_i N_i^d(c)$$

$$= \sum_i \frac{N_i(c)}{N(c)} \frac{N_i^d(c)}{N_i(c)}$$

$$= \sum_i \lambda_i(c) s_i^d(c),$$

where $N(c)$ is total employment in city c, $N_i^d(c)$ is the number of dropouts employed in industry i in city c, $N_i(c)$ is total employment in industry i in city c, $\lambda_i(c) \equiv N_i(c)/N(c)$ is the employment share of industry i in city c, and $s_i^d(c) = N_i^d(c)/N_i(c)$ is the share of dropout workers in industry i in city c. It follows that the gap between $s^d(c)$ and the national average fraction of dropouts, s^d, can be written as the sum of a "between industry component" B representing shifts in the relative employment shares of different industries in the city, a "within industry component" W, representing shifts in the relative fraction of dropout workers in each industry, and an interaction component I:

(2)
$$s^d(c) - s^d = B(c) + W(c) + I(c),$$

where

$$B(c) = \sum_i s_i^d[\lambda_i(c) - \lambda_i]$$

$$W(c) = \sum_i \lambda_i[s_i^d(c) - s_i^d]$$

$$I(c) = \sum_i [\lambda_i(c) - \lambda_i] \times [s_i^d(c) - s_i^d].$$

Under the idealized conditions of the Heckscher-Olin model, *all* of the variation in the share of dropout labor across cities can be absorbed by expansion or contraction of high-dropout-intensity industries (i.e., via the $B[c]$ term), with no city-level variation in relative wages or the dropout intensity of any particular industry.[22]

We use 2000 Census data on employment classified by three-digit industry to compute the terms in equation (2) for each of 150 larger MSAs. We then performed a series of cross-city regressions of the form:

(3a)
$$B(c) = a_B + b_B[s^d(c) - s^d] + e_B(c)$$

(3b)
$$W(c) = a_w + b_W[s^d(c) - s^d] + e_w(c)$$

(3c)
$$I(c) = a_I + b_I[s^d(c) - s^d] + e_I(c).$$

Because equation (2) holds as an identity, the coefficients b_B, b_W, and b_I sum to 1. A strict version of the HO model implies $b_B = 1$.

Figure 6.11 plots the between-industry component $B(c)$ against the excess fraction of dropouts in each of the 150 larger MSAs. For reference, note that if changing industry structure accounted for the absorption of dropouts in cities with high dropout shares the points would lie along a line with slope 1. Although the points suggest an upward-sloping relationship, the slope is relatively modest, suggesting that changing industry structure

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22. These conditions include infinitely elastic supplies of capital, perfectly integrated product markets, and the existence of at least one industry that produces a tradeable good or service that has a dropout intensity that exceeds the maximum dropout share in any city.
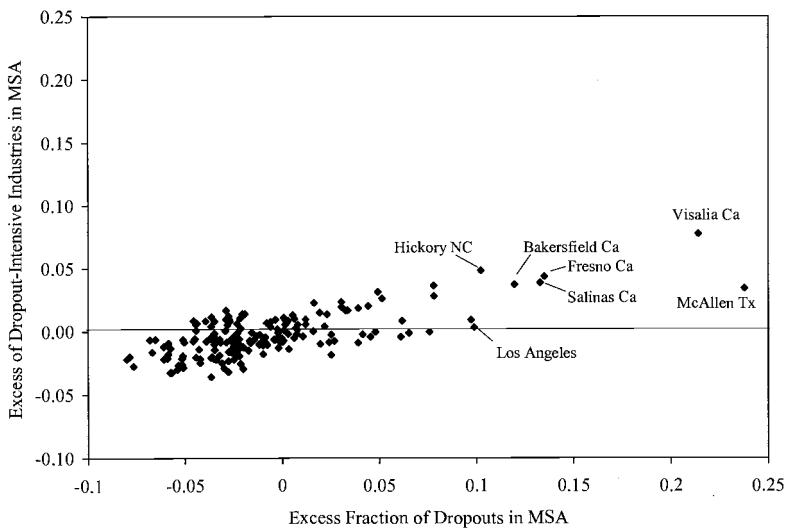
Fig. 6.11   Contribution of between-industry component to absorption of dropouts

accounts for only a small share of the absorption of dropouts. Indeed, the OLS estimate of b_B, reported in the first column of table 6.6, is 0.22, and is significantly below 1. By contrast, figure 6.12 plots the within-industry component $W(C)$ against the excess fraction of dropouts in each city. This component is more highly correlated with the dropout share, and many of the city observations are tightly clustered along the 45-degree line. The estimate of b_w, shown in column (2) of table 6.6, is 0.76. Though not shown in a figure, the interaction terms are relatively small and essentially uncorrelated with differences across cities in the share of dropout workers. Consistent with this, the estimate of b_I in column (3) of table 6.4 is 0.02 (with a very small R-squared $= 0.03$).

The MSAs that show some evidence of significant between-industry adjustment are labeled in figures 6.11 and 6.12. Interestingly, most of these MSAs represent counties in California with substantial agricultural employment.[23] The framework of equation (2) can be used to examine the contribution of specific industries to the absorption of local supplies of dropout labor. The contribution of industry i to the between-industry effect is $s_i^d[\lambda_i(c)-\lambda_i]$, which is the excess employment share of the industry in city c relative to its national average share, multiplied by the average dropout intensity of the industry. Columns (4)–(6) of table 6.6 show estimates of models similar to equation (3a), focusing on the absorption contributions of agriculture, textiles, apparel and footwear industries, and a set of

23. Hickory, North Carolina is an exception. This city is a major center for the furniture industry.

Table 6.6 **Regression models measuring cross-city absorption of excess dropout workers or Mexican immigrants**

	Absorption of excess fraction of dropout workers						Absorption of excess fraction of Mexican immigrants					
				Sector-specific absorption						Sector-specific absorption		
	Between industry (1)	Within industry (2)	Interaction (3)	Agriculture (4)	Textiles apparel (5)	Low-skill services (6)	Between industry (7)	Within industry (8)	Interaction (9)	Agriculture (10)	Textiles apparel (11)	Low-skill services (12)
Excess fraction of dropouts or Mexican immigrants	0.22 (0.02)	0.76 (0.02)	0.02 (0.01)	0.09 (0.02)	0.05 (0.01)	0.03 (0.01)	0.06 (0.01)	0.92 (0.01)	0.01 (0.01)	0.04 (0.01)	0.01 (0.00)	0.01 (0.00)
R^2	0.37	0.84	0.03	0.17	0.24	0.33	0.25	0.96	0.01	0.14	0.13	0.41

Notes: All models estimated across 150 larger cities, using 264 industry cells per city. Regressions are weighted by city size.
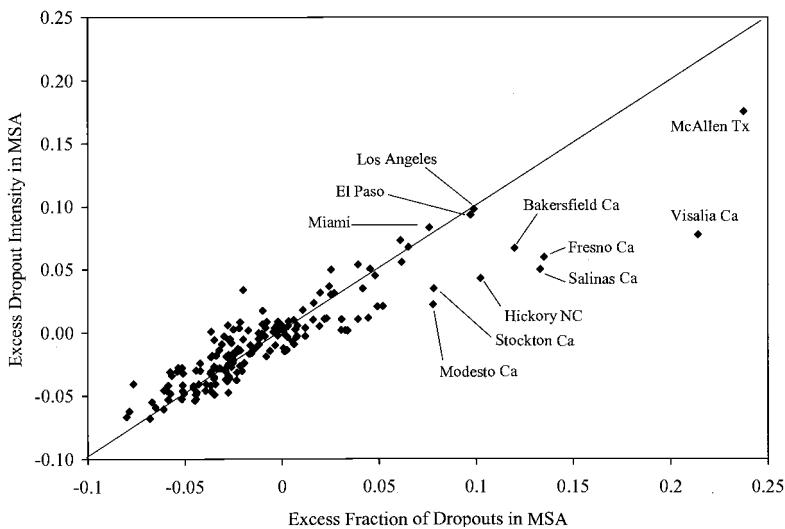
Fig. 6.12   Contribution of within-industry component to absorption of dropouts

low-skilled service industries.[24] The estimates suggest that these three industry clusters account for most of the between industry effect observed in column (1): agriculture alone accounts for nearly one-half.

Figure 6.13 plots the between-industry component of absorption of dropout labor in different cities *excluding agriculture,* while figure 6.14 shows the absorption contributions of agriculture industries and textiles and apparel industries. Overall, though there is some evidence that textiles and apparel manufacturing tends to cluster in cities with moderately high dropout shares and that agricultural employment is higher in cities with very high dropout shares, the results in table 6.6 and figure 6.14 suggest that most of the absorption of unskilled labor across cities occurs within industries rather than between.

Similar conclusions were reached by Lewis (2003), who examined changes in the absorption of workers in four education groups over the 1980–1990 period. Lewis used census data to estimate first-differenced versions of equation (3a) for each skill group. He also compared OLS estimates to IV estimates that used immigrant inflows based on historical immigration patterns as instruments for the changes in the relative shares of each skill group.[25] A potential advantage of a first-differenced approach is

24. We include textiles, apparel, knitting mills, footwear, and leather industries as apparel and the following as "low-skilled services": building services, landscaping services, carwashes, landscaping, dry cleaning and laundry services, private household services, and other personal services.

25. One difference is that Lewis regresses the between-industry effects on the population share of the skill group in the local labor market, rather than the employment share.
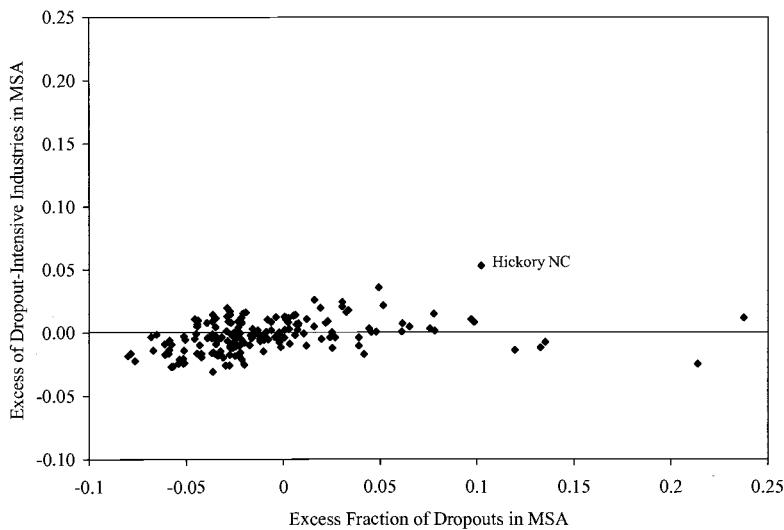
Fig. 6.13   Contribution of between-industry component to absorption of dropouts, excluding agriculture
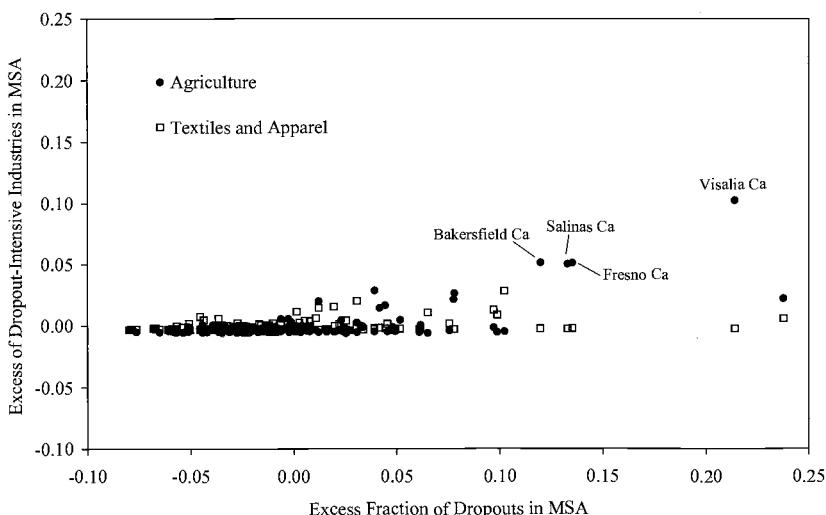

Fig. 6.14   Contribution of between-industry component to absorption of dropouts, agriculture and textiles or apparel industries

that it eliminates any MSA-specific factors that are constant over time and affect the attractiveness of the MSA to different industries (such as the amount of agricultural land available). Consistent with the pure cross-sectional results here, however, Lewis finds that changes in the scale of different industries are only weakly related to changes in the relative supply

of different skill groups. Lewis's estimates of b_B for manufacturing industries (which can readily expand their sales beyond the local market) are very close to 0, while his estimates for all industries range from 0 to 0.08. He also reports parallel specifications in which the dependent variable is the within-industry relative employment term. These are much more strongly correlated with relative population growth, accounting for 90 percent of the adjustment to skill-group specific relative supply shocks.

As a final exercise, we conducted a parallel analysis focusing on the absorption of Mexican immigrants. The relation between the within-industry absorption component and the share of Mexican workers in the local labor market is plotted in figure 6.15, while regression models similar to the models for dropout workers are reported in columns (7)–(12) of table 6.6. The results reinforce our conclusions based on an analysis of total dropout labor. In particular, over 90 percent of the adjustment to differences in the local availability of Mexican labor is explained by differences in the utilization of Mexican labor *within* three-digit industries. Surprisingly, there is almost no evidence that availability of Mexican immigrant labor stimulates low-skill service employment.

Taken as a whole, the results in this section suggest that HO-style changes in industry structure play a relatively small role in explaining how cities have been able to absorb inflows of relatively unskilled Mexican immigrants over the 1990s. Contrary to our initial expectations, most of the inflows appear to be absorbed by city-specific–within-industry increases in use of unskilled labor.
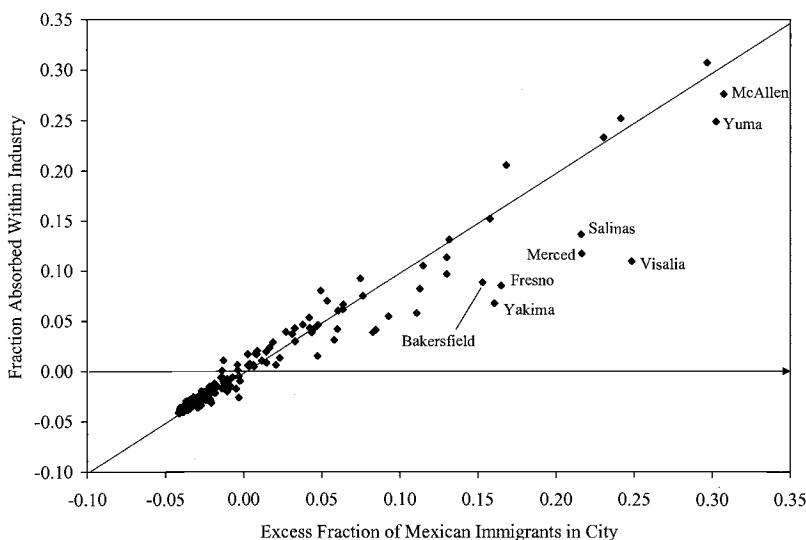

Fig. 6.15   Within-industry component of absorption of Mexican immigrant workers

6.3.3   Relative Wage Adjustments

The observation that variation in the relative supply of dropout labor is mainly absorbed by changes in utilization within industries points to the potential importance of relative wage adjustments in response to inflows of Mexican labor. We analyze relative wages in the framework of a conventional constant elasticity of substitution (CES) production function. The results in the last section suggest that we can ignore differences across industries and focus on a one industry model. Specifically, consider a production function for a single local output good:

$$y = \left[\sum_j (e^j N^j)^{(\sigma-1)/\sigma} \right]^{\sigma/(\sigma-1)}$$

where N^j is the number of people employed in skill group j, e^j is a relative productivity shock, and σ is the elasticity of substitution between labor types. Given a set of wage rates w^j for different skill groups, the relative labor demand curve between any two skill groups, say d = dropout labor and H = high school graduate labor, can be written as

$$\log\left(\frac{N^d}{N^H}\right) = -\sigma \log\left(\frac{w^d}{w^H}\right) + (\sigma-1)\log\left(\frac{e^d}{e^H}\right).$$

This equation shows that employers can be induced to increase the relative utilization of dropout labor by reducing the relative wage of dropout workers. Inverting the relative demand curve leads to a simple estimating equation that relates the relative wage gap between high school graduates and dropouts in a city to the relative supply of the two types of workers:

$$(4) \qquad \log\left(\frac{w^H}{w^d}\right) = \frac{-1}{\sigma} \log\left(\frac{N^H}{N^d}\right) - \frac{(\sigma-1)}{\sigma} \log\left(\frac{e^H}{e^d}\right).$$

As has been recognized in the immigration literature, a problem for the estimation of a model like (4) is that local relative demand shocks may raise relative wages *and* attract differential inflows of skilled versus unskilled workers. To address this concern, we consider a first-differenced version of (4) that abstracts from any permanent characteristics of a city that may affect the relative demand for less-skilled labor. We also consider IV estimates of the first-differenced model in which we use the supply-push variables (lagged Mexican immigrant densities in the city) to instrument the change in the relative supply of dropout labor in a city.

Table 6.7 presents estimation results for equation (4), based on data for 145 larger MSAs. We measure the dependent variable as the difference between regression-adjusted mean log wages for native male workers in a city with exactly twelve years of schooling and those with less than twelve years of schooling. Following the recent inequality literature (e.g., Katz and Murphy 1992) we measure the supply of high school workers in a city by

Table 6.7   **Regression models for wage gap between high school and dropout native male workers**

	Estimated by OLS				Estimated by IV—Change: 1990–2000	
			Change: 1990–2000			
	2000	1990				
	(1)	(2)	(3)	(4)	(5)	(6)
Log relative supply (high school vs. dropout labor)	0.01 (0.01)	–0.03 (0.01)	–0.04 (0.04)	–0.05 (0.04)	0.00 (0.07)	–0.04 (0.06)
Employment growth, 1990–2000				–0.06 (0.04)		–0.01 (0.05)
F-statistics for first-stage models:						
Model for log relative supply					26.8	10.68
Model for employment growth						19.02
R^2	0.00	0.04	0.01	0.02	0.00	0.00

Notes: All models estimated on sample of 145 larger cities with census data for 1980–2000 and matching employment data from county business patterns for 1982–2000. Dependent variable is gap between regression adjusted mean log wage of high school male natives in city and regression adjusted mean log wage of dropout male natives in city. Models are estimated by weighted OLS or IV using 1990 population counts of working age adults as weights. Instruments in column (5) are fraction Mexican immigrants in adult population of city in 1980 and 1990. Instruments in column (6) are fraction of Mexican immigrants in adult population in 1980 and 1990 and log of city-level employment in 1982–1990.

the number of people with a high school diploma, plus half of the number who have between thirteen and fifteen years of completed schooling. We similarly measure the supply of dropout workers as a simple count of the number with less than a high school education. The models are estimated by weighed OLS and IV, using 1990 population counts as weights.

The results for the OLS models in columns (1)–(3) suggest that there is not a large or statistically significant relationship between the relative wages of high school dropouts and their relative supply in different cities, although the point estimate of the relative supply effect in the first-differenced model is negative. We also consider a specification in column (4) that adds employment growth in the city as an additional explanatory variable. This has a modest negative effect on the wage gap, suggesting that relative wages of dropouts are higher in rapidly growing cities, though the coefficient is not significant at conventional levels. Adding this variable has little impact on the estimated supply effect.

The IV estimates in columns (5) and (6) use the shares of Mexican immigrants in the city in 1980 and 1990 as instruments for the change in the log relative supply of high school versus dropout labor. Before discussing these results, it is instructive to look at the data in figures 6.16 and 6.17,
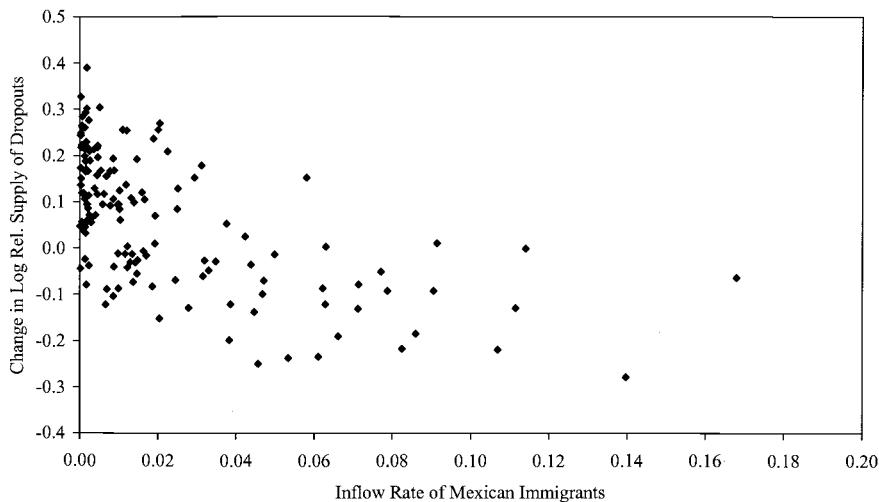
Fig. 6.16   Inflow rate of Mexican immigrants and change in relative supply of dropout labor

which illustrate the relationship between inflows of new Mexican immigrants to a city and the relative supply (figure 6.16) and relative wages (figure 6.17) of dropout labor. Figure 6.16 establishes that there is a strong impact of Mexican inflows on the relative supply of dropout versus high school labor. Given the models in table 6.4 suggesting that 75 percent or more of the variation in Mexican inflows can be explained by supply-push factors, it is clear that our IV strategy has a powerful first stage. (Indeed, the *F*-statistic for the first stage underlying the results in column [5] of table 6.7 is 26.8, with 2 and 142 degrees of freedom.) Figure 6.17, on the other hand, suggests that there is not much correlation between high school-dropout wage gap and the inflow rate of Mexican immigrants. The overall scatter of the points is slightly positively sloping (consistent with the idea that an increase in the relative supply of dropouts lowers their relative wages), but close inspection suggests that only a handful of points contribute to the slope.

The simple IV specification in column (5) of table 6.7 yields an estimate of the effect of relative supply that is somewhat less precise than the corresponding OLS model but no more negative in magnitude. The same conclusion emerges from the model in column (6), in which we treat both the change in relative supply and employment growth as endogenous.[26] It does

26. In this model, both the change in relative supply and employment growth from 1990 to 2000 are treated as endogenous, and the fractions of Mexican immigrants in the city in 1980 and 1990 and the log of employment in 1984–1990 are used as instrumental variables. The
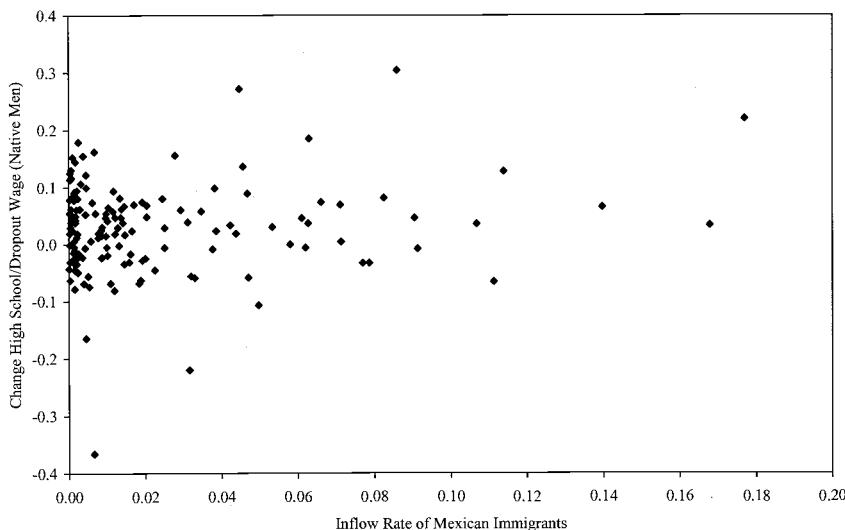
Fig. 6.17   Inflow rate of Mexican immigrants and change in relative wage of native male dropouts

not appear that increasing supplies of dropout labor arising from the predictable component of inflows of Mexican immigrants have much effect on the relative wage structure in a city.

We have also estimated a number of variants of the models in table 6.7. In one variant, we added a control for the change in the relative number of college versus high school-educated workers to the first-differenced specification in column (4). This variable has a marginally significant positive effect on the high school dropout wage gap (coefficient = 0.15, standard error = 0.07), but its addition does not have any impact on the coefficient of the variable measuring the relative supply of dropouts or on the employment growth effect. We also estimated the models using unweighted OLS and IV. The coefficient estimates from the unweighted models are somewhat less precise but show a similar pattern to the results in table 6.7. For example, the estimated relative supply effect from the first-differenced specification in column (4) is –0.07 (with a standard error of 0.05). Finally, we considered a specification in which the supply of high school workers was narrowly defined to include only those with exactly twelve years of schooling. This leads to a slightly bigger coefficient on the relative supply variable. For example, the estimate corresponding to the specification in column (4) is –0.06, with a standard error of 0.04. Overall, there is not

F-statistic for the first-stage model explaining the change in relative supply is 10.68 (with 11 and 133 degrees of freedom). The F-statistic for the first-stage model explaining the change in employment is 19.02 (with 11 and 133 degrees of freedom).

much evidence that the relative supply of dropout labor in a city has much impact on dropout relative wages.

6.3.4   Interpretation

Our findings with respect to the impacts of Mexican immigration present a puzzle. Inflows of Mexican immigrants appear to raise the relative supply of low-education labor in a city. Contrary to a simple trade-style model, however, shifts in the relative supply of low-education labor across cities do not lead to systematic expansions or contractions in dropout-intensive industries. Rather, most of the variation in the relative supply of dropout labor is absorbed by changes in dropout intensity within narrowly defined industries. Even more surprisingly, differences in dropout intensity of employment do not seem to be strongly related to the relative wages of dropout workers. Thus, it is hard to explain the variation in dropout intensity across cities as variation along a relative demand curve.

We believe there are a number of possible explanations for these findings. One is that high school dropouts are highly substitutable with high-school educated workers. Under this assumption, the share of dropouts relative to high school graduates employed in a given city will vary with local supply, but the relative wage gap between the two groups will be roughly constant. The near-perfect substitutes assumption is consistent with the fact that the aggregate wage gap between high school graduates and dropouts has been constant since 1980 (Card 2005). It is also potentially consistent with the very imprecise estimates of the inverse elasticity of substitution across education groups obtained by Borjas (2003) and Borjas and Katz (2005) using national data from the past four decades.[27] If dropouts and high school graduates are close to perfect substitutes, Mexican immigration may be depressing the relative wages of *all* workers with low and medium levels of education (e.g., up to fourteen years of schooling) relative to college graduates. Nevertheless, the proportional impact of Mexican inflows on the relative supply of workers with up to fourteen years of schooling is considerably smaller than their impacts on the relative supply of dropout labor, so if this hypothesis is true, concerns over the negative impacts of Mexican immigrants on low-wage natives are overstated.

A second possibility is that local industry structure responds to relative factor supplies as predicted by the HO model but that changes occur *within* narrowly defined industries (i.e., below the three-digit industry level). For example, if an industry consists of subsectors that use different relative

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27. For example, the estimate of the inverse elasticity of substitution across four education groups (including dropouts, high school graduates, people with some college, and people with a bachelor's degree or more) obtained by Borjas and Katz (2005) using a nested CES structure is 0.41 with a standard error of 0.31. (Revised estimate and standard error reported in personal communication from Larry Katz.) This relatively imprecise estimate does not rule out perfect substitution across different education groups.

fractions of dropout workers, then the relative supply of dropout workers in a city may determine the relative size of the dropout-intensive subsector, with little or no affect on the overall size of the combined industry. This hypothesis is observationally equivalent to the model proposed by Beaudry and Green (2005) in which output is produced by two coexistent technologies, one of which is relatively more intensive in low-skilled labor. This class of models may be useful in describing certain industries but seems less appealing for other industries, like locally traded services.

A final (closely related) hypothesis is that employers adapt to the relative supply of different skill groups in their local market without the signals of relative wage changes. Acemoglu's (1998) model of endogenous technological change, for example, suggests that firms will innovate in a direction to take advantage of more readily available factors, even in the absence of relative wage changes. Lewis (2004) presents some direct evidence for an endogenous technological change mechanism, using data on the number of advanced technologies adopted by manufacturing plants in the late 1980s and early 1990s. He finds that controlling for very detailed (four-digit) industry effects, the adoption of advanced technologies by individual plants is significantly slowed by the presence of a greater relative supply of unskilled labor in the local labor market. More work is needed to understand how firms choose which technologies to use and whether the choice is influenced by the relative availability of different skill groups, particularly low-skilled immigrants.

6.4   Conclusions

Mexicans are the largest single group of immigrants in the United States, representing about one-third of all immigrants and more than 4 percent of the country's working-age population. Until the last decade, Mexican immigrants were geographically clustered in a relatively small number of cities. In 1990, nearly one-half of all working-age Mexicans were living in just five U.S. cities, and 70 percent were living in only fifteen cities. During the 1990s, however, arrivals from Mexico established sizeable immigrant communities in many new cities, including Atlanta, Denver, Portland, and Raleigh-Durham. These immigrants are changing the face of the new destination cities and setting the stage for many years of future inflows.

In this paper we present some simple evidence on the causes and consequences of the widening geographic diffusion of Mexican immigrants. A combination of demand-pull and supply-push factors explains 85 percent of the variation across major cities in the rate of Mexican inflows during the 1990s and helps illuminate the single most important trend in the destination choices of new Mexican immigrants—the move away from Los Angeles.

Like their predecessors, recent Mexican immigrants have relatively low

levels of education. We show that inflows of Mexican immigrants lead to systematic shifts in the relative supply of low-education labor in a city, opening up the question of how different local labor markets are adopting to substantial differences in relative supply. One possibility—suggested by the conventional Heckscher Olin model of international trade—is that these differences are accommodated by shifts in industry composition. Despite the theoretical appeal of this hypothesis, we find it has limited empirical relevance: most of the differences across cities in the relative supply of low-education labor (or Mexican labor) are absorbed by changes in skill intensity *within* narrow industries. Such adjustments could be readily explained if Mexican immigrant inflows had large effects on the relative wage structures of different cities. As has been found in previous studies of the local impacts of immigration, however, our analysis suggests that relative wage adjustments are small. Thus, we are left with the puzzle of explaining the remarkable flexibility of employment demand in different cities to local variation in supply. Given the continuing pace of Mexican immigration, the next decade should provide even more evidence on the ways that local economies adjust to shifts in relative supply.

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