The International Transmission of Local Economic Shocks Through Migrant Networks^{*}

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Abstract

Using newly validated data on geographic migration networks, we study how labor demand shocks in the United States propagate across the border with Mexico. We show that the large exogenous decline in US employment brought about by the Great Recession affected demographic and economic outcomes in Mexican communities that were highly connected to the most affected markets in the US. In the Mexican locations with strong initial ties to the hardest hit US migrant destinations, return migration increased, emigration decreased, and remittance receipt declined. These changes significantly increased local employment and hours worked, but wages were unaffected. Investment in children's education also slowed in these communities. These findings document the effects in Mexico when potential migrants lose access to a strong US labor market, providing insight into the potential impacts of stricter US migration restrictions.

JEL codes: F22, J21, J23, J61, R23

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1 Introduction

Goods trade and capital flows are well-studied economic mechanisms that integrate markets across international borders, but international migration represents another potentially equilibrating force (Chiswick and Hatton 2003). Research suggests that potential international migrants consider relative economic conditions when deciding whether and where to move, and these choices affect the size and composition of the labor force in source and destination communities (Hanson and Spilimbergo 1999, Borjas 2001, Cadena 2013, Cadena and Kovak 2016). This earnings-maximizing behavior implies that local labor market conditions in potential destinations will affect demographic and economic outcomes in sending locations by changing both migration choices and the remittance behavior of existing migrants.

In this paper, we study how changes in US labor demand affect migration, demographic, and economic outcomes in migration-network-connected communities in Mexico. Changes in US labor demand should have important consequences in Mexico, as 98 percent of Mexicans living abroad are in the US and approximately 10 percent of the Mexican-born population lives in the United States.¹ We show that US labor demand declines during the Great Recession affect outcomes in Mexican sending areas with strong ties to the hardest-hit US local labor markets. This focus on sending communities contrasts with much of the literature on the economics of Mexico-US migration, which more often evaluates the impacts of Mexican migration on US destination markets.²

To motivate our empirical analysis, we derive a reduced-form estimating equation and shock measure from a simple location choice model, which shows how to leverage two key sources of variation: the heterogeneity across US destinations in employment declines during the Great Recession and differences in migration network connections between each Mexican source and each US destination. This type of empirical design requires detailed information on migration network connections between Mexican sources and US destinations, a challenge we overcome by using newly

¹Numbers living in the US and in other countries are available in Secretaría de Relaciones Exteriores (2015). Population numbers for 2010 and 2015 are available at https://www.inegi.org.mx/temas/estructura/.

²See National Academies of Sciences, Engineering, and Medicine (2017) for a survey of the literature on the effects of immigration on earnings, employment, and wages in destination countries, and Mishra (2014) and Elsner (2015) for surveys of the literature on the effects of emigration on wages in source countries.

validated administrative data from the *Matrícula Consular de Alta Seguridad* (MCAS) identification card program. These data allow us to measure the distribution of US commuting zones chosen by migrants from each Mexican *municipio* (similar to a US county), a much more granular level of geography than other data sources.³ The resulting empirical analysis compares the change in outcomes between source municipios whose migrants face larger and smaller effective declines in US employment due to their source location's mix of US commuting zones.

In order for this analysis to have a causal interpretation, a municipio's network-weighted US demand shock must be uncorrelated with other factors affecting its demographic and economic outcomes. This exogeneity assumption is likely to hold in part because the relevant demand shocks for each municipio occur in another country (the US) and are thus unlikely to be related to other changes in Mexican source communities. To strengthen the causal interpretation, we include Mexican state fixed effects so that we compare only geographically proximate municipios, and we allow for differential trends based on pre-existing characteristics of the source community. Further, we follow the model-motivated estimation strategy by controlling for contemporaneous changes in observable source-level characteristics such as drug-related violence and network-weighted averages of destination-level changes in local immigration enforcement policy. We also control for the possibility that the Great Recession affected Mexican outcomes through international trade by including a measure of each municipio's exposure to declining US-Mexico trade over the same time period. The results are robust to the inclusion of these controls, bolstering the interpretation of the key coefficient as the causal effect of declining US labor demand on Mexican source community outcomes.

Using Mexican Census data, we find that source communities with strong initial ties to the US destinations hardest-hit by the Recession experienced roughly 20 percent faster population growth from 2005 to 2010, driven in large part by a similar percentage increase in return migration and decrease in emigration. The change in migration also increased the share of the local workforce that is male, although the educational attainment distribution was relatively unaffected. Beyond

³Caballero, Cadena and Kovak (2018) confirm the quality and representativeness of the MCAS data by comparing it against high quality household survey data. Other papers using various versions of the MCAS data include Massey and Espinosa (1997), Albert and Monras (2022), Allen, Dobbin and Morten (2019), and Tian, Caballero and Kovak (2022).

the movement of people, we also find that households living in these Mexican sources experienced a 20 percent decline in the likelihood of receiving remittance payments from abroad.

These changes in population size and composition lead directly to a substantial increase in the size of the local labor force. Further, the loss of US remittance income creates an incentive for additional household members to enter the labor force. Using data from the Mexican Economic Census, we examine changes in municipio-level labor market outcomes. As expected, we find clear evidence of an expansion in labor supply, with source communities facing the largest declines in US employment opportunities seeing larger increases in employment and total hours worked, especially among women. We reinforce this result with descriptive evidence showing that the relationship between labor supply and US labor demand shocks appears only in households who had migrants in the US during the Great Recession period. Interestingly, we find no evidence that this expansion of labor supply led to a relative decrease in local wages. This result is not driven by changes in the composition of the local labor force and is consistent with much of the literature on the effect of immigration on host labor markets, potentially reflecting the fact that the return of migrants also increased local labor demand. We then document the effects of declining US labor demand on household investment behavior, both in durable goods and human capital. We find minimal effects on appliance ownership but find that children in the most affected communities are less likely to remain in school, especially at late primary school ages.

Taken together, these results demonstrate that migrant networks transmit changes in US local labor demand across the southern border with Mexico, leading to significant effects on a wide variety of outcomes in sending areas. Further, because we study job loss among migrants, the results show how losing access to US employment affects economic outcomes in Mexican migrant-sending communities. In addition to documenting the impacts of the US Great Recession across Mexican communities, our findings therefore provide insight into the potential impacts of proposed migration enforcement policies, such as requiring firms to check a job applicant's work authorization in an electronic database before hiring them (E-Verify).⁴

⁴Related papers document population responses to Arizona's statewide E-Verify policy, with migration results similar to what we find in this paper (Bohn, Lofstrom and Raphael 2014, Caballero et al. 2018). The Comprehensive

This paper contributes to multiple strands of literature. First, as mentioned above, many papers find that international migrants' location choices respond to local labor market conditions.⁵ Borjas (2001) simulated how foreign-born workers' location choices might equalize native workers' wages across regional labor markets in the destination country, and Cadena and Kovak (2016) empirically measure these equalizing effects, showing that a metropolitan area's local population of Mexicanborn workers with no more than a high school degree was strongly responsive to changes in local labor demand during the Great Recession. Here, we demonstrate that differential return migration to Mexico also contributes to the reallocation of immigrants across US markets, whereas previous work had provided only suggestive evidence of this channel's importance.

Second, this study expands our understanding of the role of networks in driving international migration. Larger numbers of migrants from a sending community increase the likelihood of subsequent migration by lowering migration costs, especially for those with relatively low levels of education (Winters, de Janvry and Sadoulet 2001, McKenzie and Rapoport 2007, Garip and Asad 2016). Migrant networks also affect migrants' destination locations, occupational choices, and labor market success (Munshi 2003, Edin, Fredriksson and Åslund 2003). We extend this literature by developing a tractable model showing how migrant networks lead to source communities facing differential changes in foreign labor demand from the same macroeconomic shock. The model-motivated empirical analysis demonstrates that these network connections serve to transmit local economic shocks from one side of an international border to the other.

Third, our results relate to the substantial literature examining the effects of international migration on family members who remain in the source country. As summarized nicely in Antman (2013), demand shocks at the destination are one of two commonly used instruments in this body of work.⁶ Relative to this literature, our analysis is distinct in two ways. First, in contrast to typical

Immigration Reform Bill that passed the US Senate in 2013 included a mandatory national E-Verify provision. More recently, in February 2021, Senators Romney and Cotton proposed universal E-Verify as a condition for raising the federal minimum wage to \$10 (King 2021).

⁵More generally, these results confirm the consistent finding that both initial and return migration respond to relative labor market conditions in sending and receiving communities (Wozniak 2010, McKenzie, Theoharides and Yang 2014, Abarcar 2017, Bertoli, Fernández-Huertas Moraga and Keita 2017).

⁶Examples include Antman (2011) and Cortes (2015).

studies in this literature that study increased emigration, we consider the effects of a decline in destination labor demand and thus the effects of increased return migration, deferred emigration, and a decline in remittances. Second, we consider outcomes at the municipio level rather than the household level, which allows us to use a wide array of high-quality survey and administrative data sources to measure the overall effects on local markets including any cross-household spillovers.

Finally, we extend the literature examining how destination-market policies or labor demand affect sending communities. The studies most closely related to this paper consider the effects of changes in the US environment on economic outcomes in Mexico. Caballero et al. (2018) and Allen et al. (2019) use MCAS migration network data to show that migration enforcement reduces international migration between affected sources and destinations, and Caballero (2022) uses the rollout of the Secure Communities program to show that local migration enforcement at the destination reduces school enrollment in connected sources. Multiple papers study the impact of US labor demand conditions on a variety of Mexican outcomes, such as occupational choices, entrepreneurship, inequality, and education (McKenzie and Rapoport 2007, Schnabl 2007, Fajardo, Gutierrez and Larreguy 2017, Conover, Khamis and Pearlman 2021).⁷ While largely supporting the findings of this prior work, we make multiple additional contributions. Our location choice model clarifies how to combine information on migration network connections, variation in labor demand across US destinations, and source locations' exposure to the US labor market in an internally consistent empirical research design. We also use uniquely detailed geographic information in the US and Mexico, allowing us to measure shocks to well-defined US local labor markets and to compare outcomes among municipios within the same Mexican state, strengthening causal identification.

Another set of closely related papers considers the effects of shocks to emigrants' earnings on sending communities in contexts other than the US and Mexico. Yang (2008), Theoharides (2018), and Khanna, Murathanoglu, Theoharides and Yang (2022) combine variation in the historical destination countries of migrants from different source communities within the Philippines

⁷In historical contexts, Kosack (2021) studies the effect of differential access to the US Bracero program on Mexican human capital investment, and Brum (2019) studies the effects of economic shocks in US counties on migration from Italian municipalities.

with destination-level shocks. Gröger (2021) uses a similar methodology focusing on Vietnamese households with migrants in different destination countries at the onset of the Great Recession. Our research design is closely related and reaches similar conclusions in the Mexico-US context, which has the advantage of using variation in labor demand across migrant destinations within the same country. Because each municipio sends migrants almost exclusively to the US, our analysis is robust to other nationwide changes to the attractiveness of living abroad, including immigration enforcement, visa availability, or exchange rates. Dinkelman and Mariotti (2016) and Dinkelman, Kumchulesi and Mariotti (2022) use a particularly compelling research design leveraging exogenous changes in emigration restrictions in Malawi to yield highly credible estimates of the causal effects of remittances on educational attainment, capital accumulation, and the structure of rural labor markets in migrant sources. In the absence of such policy changes in the Mexican context, our approach combines shocks across migrant destinations with persistent geographic migrant networks to generate similar variation across migrant sources in access to higher-paying foreign labor markets.

The remainder of the paper is organized as follows. Section 2 introduces our dataset and demonstrates that historical settlement patterns led to substantial variation in how Mexican source municipios experienced the US Great Recession. Section 3 develops the location choice model that leads to our estimation strategy. Section 4 discusses the US and Mexican data sources we compile to execute our analysis. Section 5 shows that larger negative labor demand shocks in the US led to increased return migration, decreased emigration, and a decline in the share of households receiving remittances. Section 6 then demonstrates that these changes increased local employment without decreasing average wages, while reducing school enrollment among children. Section 7 concludes.

2 Motivation and Context

Mexican source communities face different changes in US labor demand for two reasons: 1) changes in local labor demand were different across US local labor markets, and 2) migrant-sending communities in Mexico have historical ties to different sets of destinations within the US. In this section, we provide descriptive evidence documenting these two key facts.

2.1 Geographic Variation in Job Loss During the Great Recession

Identifying labor demand shocks is challenging because observed changes in employment and earnings normally reflect changes in both labor demand and labor supply. To overcome this challenge, we take advantage of the unique environment during the Great Recession. Beginning in December 2007 and lasting through June 2009, this decline in economic output was marked by a more than five-percentage-point drop in the prime age employment-to-population ratio. While there was a dramatic reduction in hiring and a large increase in layoffs, wages did not fall substantially (Rothstein 2012, Daly, Hobijn and Wiles 2012). This pattern suggests that the labor market adjusted primarily along the employment margin rather than through wage reductions. Given downward-rigid wages, one can measure local labor demand shocks over the Great Recession period (2006-2010) using only changes in payroll employment (Cadena and Kovak 2016, Clemens 2022).

Figure 1 shows the substantial variation in employment changes from 2006-2010 across US local labor markets.⁸ This map uses data from the County Business Patterns (CBP) and the American Community Surveys (ACS) to show changes in employment in US Commuting Zones (CZ), which define destination labor markets throughout the paper.⁹ We account for the industry mix of Mexican workers' US employment by measuring the relevant employment change in each commuting zone das $\sum_{i} \frac{Emp_{id}^M}{Emp_{id}^{2010} - Emp_{id}^{2006}}$, where Emp_{id}^{year} is employment in industry i in destination commuting zone d and $\frac{Emp_{id}^M}{Emp_{id}^M}$ is the share of Mexican-born workers in commuting zone d working in industry i in 2006.¹⁰ This measure accounts for the fact that Mexican-born workers are disproportionately represented in industries that are especially sensitive to the business cycle, such as construction.¹¹ Appendix C.1 provides descriptive statistics detailing the sources of spatial variation in this measure: variation in the pre-recession industry mix of employment among Mexican-born workers and

⁸Multiple factors contributed to the substantial spatial variation in local employment declines, including variation in ex-ante household indebtedness (Mian and Sufi 2014) and in the magnitude of the pre-Recession housing boom (Charles, Hurst and Notowidigdo 2016).

⁹See Appendix Section B.3 for details on the CZ definition.

¹⁰Appendix Section B.3 explains how we combine CBP and ACS data in this measure.

¹¹The main results are qualitatively similar, however, when using unweighted CZ-level employment declines.

spatial variation in industry-specific shocks. As Figure 1 shows, although most commuting zones experienced a decline in employment, there was substantial variation, with a 17 percent decline at the 25th percentile and no change at the 75th percentile.

2.2 Matrículas Consulares de Alta Seguridad

In addition to this spatial variation in US labor demand, we leverage variation in the destinations historically chosen by migrants from different Mexican source communities. We measure source-destination connections using administrative tabulations from Mexico's *Matrícula Consular de Alta Seguridad* (MCAS) program, in which Mexican consulates issue identity cards to Mexican-born individuals living in the US. The cards, which provide a secure form of identification and verified current residence for banking and other purposes, are issued primarily to those without authorization to live and work in the US and who therefore cannot access other forms of identification. Measuring connections between sending and receiving communities using the choices of unauthorized migrants is not a concern in our context for two main reasons. First, more than 90 percent of moves between Mexico and the US occur among unauthorized migrants during our sample period of 2006-2010 (authors' calculations using Mexican Migration Project data). Second, Caballero et al. (2018) show that the migration patterns in the MCAS data accurately reflect those of the broader Mexican-born population living in the US, irrespective of legal status.

To examine the variation in US destinations for migrants from different source municipios, we calculate $\frac{m_{sd}}{\sum_{d'} m_{sd'}}$, i.e. the share of card recipients born in source municipio *s* who settled in destination commuting zone *d* in 2006—the first year the MCAS tabulations are available. To calculate these shares, we use a customized extract from the MCAS administrative database that captures Mexican-born individuals' birthplace and county of residence in the US, which we aggregated to the CZ level.¹² The publicly available tabulations used in Caballero et al. (2018), in contrast, report only Mexican migrants' *state* of residence in the US.

¹²See Appendix B for details on matching geographic locations in the MCAS extract to municipios and counties. Special thanks to Melanie Morten for providing the specific version of the extract used in this study.

2.3 Migrants from Nearby Sources Settle in Distinct Destinations

As a motivating example of the variation in migrant destinations, Figure 2 compares the destination distributions for two Mexican source municipios in the state of Guanajuato: Dolores Hidalgo and Jaral del Progreso. Our empirical analysis controls for Mexican state fixed effects, so we are especially interested in within-Mexican-state differences in chosen destinations. Despite these two source communities' close proximity, there are large differences in the US destinations selected. Migrants from Dolores Hidalgo tend to move to the main cities of Texas, while migrants from Jaral del Progreso concentrate in Chicago, the largest cities of California, and other cities in the Southwest. As shown in Figure 1, the Texas cities faced particularly mild labor demand declines during the Great Recession, while southern California and the Southwest saw larger negative shocks. Thus, migrants from Jaral del Progreso experienced a larger effective decline in US labor demand compared to migrants from Dolores Hidalgo.

3 Theoretical Framework and Research Design

To formalize the idea that potential migrants from different Mexican source locations experienced the US Great Recession differently, we use a location choice model in which Mexican-born individuals choose to live in Mexico or in one of many potential US destinations. Potential migrants benefit from living alongside others from their place of birth, a model feature motivated by the variation in geographic migrant networks documented in Figure 2. We use comparative statics from the model to motivate our estimating equation and to clarify the set of potential confounding variables that must be controlled for to identify the causal effect of US labor demand shocks on Mexican outcomes.

3.1 Location Choice Model

An individual j from Mexican source community s may choose to live in any destination d, including their municipio of birth s or any of the potential US destination commuting zones. For simplicity, we assume costless migration and ignore internal migration within Mexico. Individual j's utility from choosing destination d depends on three things: the common-across-sources value v_d of living in that location, a network component reflecting the presence of prior migrants from the potential migrant's source n_{sd} , and an iid type-I extreme value shock η_{jsd} .

$$u_{jsd} = \alpha v_d + n_{sd} + \eta_{jsd} \tag{1}$$

The probability that a person born in s chooses to live in d is then

$$P_s(d) = \frac{\exp\left(\alpha v_d + n_{sd}\right)}{\sum_{d'} \exp\left(\alpha v_{d'} + n_{sd'}\right)}.$$
(2)

We examine how population growth in each source municipio is affected by a set of shocks to the value of locating in the various potential destinations. Let M_s be the number of people born in Mexican source s, and let M_{sd} be the number of people born in source s living in destination d. The population residing in s is therefore $M_{ss} = M_s P_s(s)$, i.e. the number of people born in smultiplied by the probability that a person born in s stays in that location. Assume that the total number of people born in source s (M_s) is invariant to changes in destination values (i.e. shocks do not affect mortality). As shown in Appendix A, taking the total derivative of M_{ss} with respect to changes in values v_d for all possible destinations and evaluating the changes in choice probabilities using (2) yields the following expression relating the proportional change in source s population to the shocks to the value of living in each potential location:

$$\frac{dM_{ss}}{M_{ss}} = \alpha \xi_s \left[dv_s - \sum_{d \neq s} \varphi_{sd} dv_d \right]$$
(3)
where $\xi_s \equiv (1 - P_s(s))$ and $\varphi_{sd} \equiv \frac{P_s(d)}{1 - P_s(s)}$

This expression is intuitive. The term ξ_s is the share of people from source s who had chosen to live in the US prior to the shock—baseline exposure to the US labor market. The first term in square brackets is the change in the value of living in the source community. As its own conditions improve, it attracts more residents, and this effect is larger when there are more residents abroad to attract. The second term in square brackets captures the effects of changing conditions in the US as mediated through the migrant network $(n_{sd} \text{ in } (1))$. This term is a proper weighted average of shocks in US destinations, where the weights, φ_{sd} , reflect the baseline distribution of migrants from s across US destinations $(d \neq s)$. As conditions in the US labor markets to which source s has existing network connections improve, more people leave s for the US.

To study how changes in labor demand across US destinations affected demographic and economic outcomes in Mexican source communities, we parameterize the value of living in each US destination commuting zone. The common value of living in US location $d \ (\neq s)$ depends on expected earnings and other factors such that

$$v_d = w_d \cdot Pr(emp_d) + \Gamma_d,\tag{4}$$

where w_d is the real wage, $Pr(emp_d)$ is the probability of employment, and Γ_d captures other features affecting the attractiveness of destination d. We take the change in (4) holding w_d fixed based on the wage rigidity observed during the Great Recession (discussed in Section 2), and plug it into (3), yielding the following expression.

$$\frac{dM_{ss}}{M_{ss}} = \alpha\xi_s dv_s - \alpha\xi_s \left[\sum_{d \neq s} \varphi_{sd} w_d \ dPr(emp_d) \right] + \alpha\xi_s \sum_{d \neq s} \varphi_{sd} \ d\Gamma_d + \nu_s \tag{5}$$

This expression forms the basis of our reduced form estimation equation, which relates source municipio population growth to changes in the attractiveness of the source community (dv_s) , changes in employment probabilities across US destinations $(dPr(emp_d))$, and other changes affecting the attractiveness of particular destinations within the U.S $(d\Gamma_d)$.¹³

¹³Note that we normalize $d\Gamma_s = 0$, so the $d\Gamma_d$ for $d \neq s$ reflect changes in the attractiveness of US destination d relative to staying in Mexico.

3.2 Estimating Equation

To empirically operationalize (5) we must first construct an observable measure of changes in expected earnings. We assume that i) the employment probability facing Mexican-born residents of d is given by the employment to population ratio among the Mexican-born population, ii) baseline expected earnings are equal across US destinations, and iii) job losses in a given industry and commuting zone are allocated proportionately to Mexican-born and US-born workers.

Given these assumptions (see Appendix A),

$$\sum_{d \neq s} \varphi_{sd} w_d \ dPr(emp_d) = \delta \sum_{d \neq s} \varphi_{sd} \sum_i \frac{Emp_{id}^M}{Emp_d^M} \cdot \frac{d \ Emp_{id}}{Emp_{id}},\tag{6}$$

where δ is the baseline expected US earnings for Mexican workers, assumed constant across destinations, Emp_{id} is employment in industry *i* in destination *d*, Emp_{id}^M is Mexican employment in *i* and *d*, and Emp_d^M is overall Mexican employment in *d*. In Appendix A, we show that, under the additional assumption that wages are constant across locations, this shock to expected earnings can be interpreted as the wage times the number of US jobs lost per migrant.¹⁴

In addition to US employment shocks, (5) shows that source-municipio population growth is also affected by changes in amenities in the source municipio (dv_s) or US destinations $(d\Gamma_d)$. We account for changes in source-municipio amenities in three ways. First, we include Mexican-state (*entidad federal*) fixed effects, $\phi_{e(s)}$ to account for changes in the value of living in one's home community that are common to municipios within the same Mexican state. Second, we show that the estimates are robust to controlling for a vector ΔX_s of changes in municipio-level characteristics, including changes in local homicide rates and trade shocks. Third, we control for pre-Recession differences in outcome growth to account for any unobserved persistent changes in source-level amenities.

Equation (5) shows that changes in destinations' non-earnings amenities also enter the expression in a weighted average, where the weights, φ_{sd} , are identical to those in the US employment shock measure. We therefore control for weighted averages of changes in CZ-level characteristics, ΔX_d ,

¹⁴Thanks to Craig McIntosh for suggesting this interpretation.

including local immigrant enforcement measures and employment policies. Our results are robust to including or excluding these various source- and destination-level controls.¹⁵

Finally, note that all of the terms on the right side of (5) are proportional to the source's exposure to the US labor market, ξ_s . For expositional clarity and to aid in interpreting the associated regressions, we divide the entire expression by ξ_s . This approach turns an estimating equation with heterogeneous effects by source s (5) into a version with homogeneous effects. We also plug in the controls just discussed and the observable US employment shock in (6), and replace the parameters α and δ with reduced-form regression coefficients, β , Λ , and Π , yielding the following estimating equation,

$$\frac{1}{\xi_s} \Delta y_s = \beta \left[\sum_{d \neq s} \varphi_{sd} \sum_i \frac{Emp_{id}^M}{Emp_d^M} \frac{d \ Emp_{id}}{Emp_{id}} \right] + \phi_{e(s)} + \Lambda \Delta \mathbf{X_s} + \Pi \sum_{d \neq s} \varphi_{sd} \Delta X_d + \varepsilon_s, \tag{7}$$

where $\varepsilon_s = \nu_s/\xi_s$, and Δy_s indicates a change in a generic source-level outcome. Note that the Mexican state fixed effects, $\phi_{e(s)}$, subsume the standard intercept term. This equation relates the exposure-normalized change in outcome in municipio s to the change in US employment faced by migrants from that destination.¹⁶ Incorporating the measure of exposure ξ_s in (7) also resolves the "incomplete shares problem" emphasized by Borusyak, Hull and Jaravel (2022), as exposure reflects the overall share of the source municipio's population in the US labor market.

Because dividing the dependent variable by ξ_s may introduce heteroskedasticity, we use feasible GLS weighting to improve the efficiency of our estimates, following Wooldridge (2013) Section 8.4. We present two sets of standard error estimates. First, we report standard errors clustered at the Mexican commuting-zone level when reporting any regression coefficient.¹⁷ Second, we account for cross-municipio correlation in our shift-share shocks by calculating standard errors following Borusyak et al. (2022), shown in square brackets.

¹⁵Table 1 presents specifications with and without the various controls, and Appendix C.2 shows specifications with subsets of controls for the remaining outcomes.

¹⁶The municipio-level US employment shock varies extensively for all values of exposure (Appendix C.3).

¹⁷See Appendix C.4 for unweighted results and Breusch-Pagan test statistics for heteroskedasticity supporting the conclusion that the weighted analysis improves efficiency. We define Mexican commuting zones following Atkin (2016), making manual adjustments for changing municipio boundaries.

Our coefficient of interest, β in (7), compares the change in outcome between municipios in the same Mexican state whose migrants faced different US employment declines during the Great Recession. To interpret this relationship as causal, there must be no unobserved variables affecting municipio outcome growth that are correlated with the municipio-specific US employment declines. This assumption will be satisfied if the shocks are exogenous as in Borusyak et al. (2022), which is plausible in our context because unobserved developments in Mexican municipios are unlikely to be related to US labor demand shocks in the municipio's historical migrant destinations. Alternatively, this assumption could be satisfied through the "exogenous shares" approach of Goldsmith-Pinkham, Sorkin and Swift (2020), which is also plausible in our setting, given that connections between Mexican municipios and US counties are often the result of historical accidents.¹⁸

Despite the plausible exogeneity of the shift-share shock variable, it remains possible that municipios' US labor demand shocks were correlated with changes in other factors that affected residents' location choices or labor market outcomes. For example, if industry mixes were similar in migrant sources and destinations, then common industry shocks could lead to spurious correlation in outcomes across countries. We therefore include additional controls for the municipio's estimated drop in export demand due to the US recession and for local drug-related violence using the functional form suggested by the model, and the results are robust to their inclusion.¹⁹

4 Data and Measurement

Throughout our analysis, we treat Mexican municipios as independent migrant source communities and US Commuting Zones (CZs), which are designed to represent integrated labor markets, as potential migrant destinations.²⁰ The US employment shock is calculated using information on the migration network and changes in US employment from before to after the Great Recession. We measure the migration network term as $\varphi_{sd} = m_{sd} / \sum_{d'} m_{sd'}$, i.e. destination d's share of MCAS

¹⁸Appendix C.5 supports this interpretation by showing baseline balance on observable demographic, educational, and labor-market characteristics across municipios with different primary migrant destinations in the US.

¹⁹Mendez (2014) provides evidence that differential ties to the US through the manufacturing sector were an important driver of spatial variation in labor market outcomes over this same period.

²⁰See Appendix B for details on variable construction. Appendix C.6 shows results for Mexican Commuting Zones.

cards issued to migrants from source s in 2006. Emp_{id} is employment in industry i and commuting zone d in 2006, and $d Emp_{id}$ is its change from 2006 to 2010. Emp_{id}^M/Emp_d^M measures the share of Mexican-born workers living in CZ d who work in industry i, which we calculate using the 2006 American Community Survey.

The exposure term, ξ_s , reflects the share of those born in a given source municipio who live in the US. Because this stock of migrants at the source-destination level, M_{sd} , is not directly observable in any data source that we are aware of, we combine 2006 ACS estimates of the stock of Mexican-born migrants living in each US destination with migrant network information from MCAS. Specifically, we apportion the 2006 Mexican-born population observed in each destination, M_d , to source municipios based the each source's share of identity cards issued to residents of that destination CZ in 2006: $M_{sd} = \left(\frac{m_{sd}}{\sum_{s'} m_{s'd}}\right) M_d \quad \forall d \neq s$. Finally, we calculate the Mexican-born population living in each source municipio, M_{ss} , using the 2005 Mexican Inter-Censal Count. The exposure for source s is then the share of people from the source living in the US: $\xi_s = \frac{\sum_{d\neq s} M_{sd}}{M_{ss} + \sum_{d\neq s} M_{sd}}$.

We examine the effects of US employment shocks on demographic and economic outcomes in Mexican municipios. We measure most outcomes using full-count tabulations from the 2005 Inter-Censal Count and 2000 and 2010 Mexican Censuses of Population. This survey timing allows us to measure key outcomes over the time period from 2005 to 2010 and to control for prior changes in outcomes from 2000 to 2005.²¹ These dependent variables include population growth, return migration, the population sex ratio, educational attainment among adults, household appliance ownership, and school attendance among children. For emigration and household remittance receipt, we use the 2010 Census and the 2000 Census because the 2005 Inter-Censal Count omits questions on these topics. Finally, we measure municipio aggregate labor earnings and aggregate hours in the 1999, 2004, and 2009 Mexican Economic Census. This data source allows us to measure changes in earnings, hours, and earnings per hour from 2004 to 2009 and pre-existing changes from 1999 to 2004.²²

 $^{^{21}}$ Table 1 and Appendix C.2 show results with and without pre-shock outcome controls. Appendix C.7 presents placebo analyses relating pre-Recession municipio outcomes to the subsequent US employment shock.

²²These pre-Recession outcome controls directly address the possibility that pre-existing population growth differed across sources facing different shocks (Monras 2020a). This approach also partly absorbs variation that might

In addition to pre-Recession outcome controls, we present specifications controlling for other municipio-level developments, including changes in the local homicide rate and changes in trade with the US. Because these controls may themselves be affected by the US employment shocks, we show that our findings are robust to including or excluding them from the analysis in Appendix C.2. We control for local homicides to capture the effects of drug-related violence using administrative data from the Mexican Statistical Office (INEGI) to calculate the number of homicides during 2005-2010 divided by the 2005 population from the Inter-Censal Count. We control for the sharp reduction in trade between Mexico and the US during the Great Recession using a weighted average of industry-level changes in trade value from the period 2001-05 to the period 2006-10, weighted by the municipio's initial industry mix of employment in 2004.²³ We also account for non-employmentbased changes in US destinations' attractiveness to potential migrants (X_d) . These controls include indicators for new state-level anti-immigrant employment legislation and indicators for new 287(g) agreements that allow local officials to enforce federal immigration law, with both variables from the immigration policy database complied by Bohn and Santillano (2017). For each of these measures, we follow (7) and calculate a weighted average of changes in the policy indicators using a source municipio's destination distribution as weights.

We limit our analysis to the source municipios for which we can accurately measure both the US employment shock and key dependent variables. Following the location-choice model in Section 3, we initially focus on municipio population growth and the contributions of decreased emigration and increased return migration. Because these dependent variables are measured as shares of the initial population, they are highly sensitive to measurement error in small-population municipios. Further, measuring a municipio's destination distribution accurately requires a sufficient number of MCAS observations. To address each of these concerns, we limit the main analysis sample to the 866 municipios that had at least 5,000 residents in the year 2005, had exposure $\xi_s > 0.066$ (the 25th percentile), and whose citizens received at least 100 MCAS cards in 2006.²⁴

confound the analysis if markets adjust slowly to prior shocks (Jaeger, Ruist and Stuhler 2019).

²³See Appendix B for details. We control for the share of employment in nontradable sectors in 2004 to address the "incomplete shares problem" (Borusyak et al. 2022).

²⁴The municipios in our analysis sample account for more than 56 percent of the working-age Mexican population

Figure 3 demonstrates the geographic variation in employment shocks experienced by each Mexican source community, controlling for Mexican-state fixed effects. We show municipios facing larger US demand declines in darker blue and municipios connected to smaller declines in lighter blue. Municipios excluded from our analysis are shown in white. The differences in US employment shocks, even for geographically proximate municipios in the same state, provide the identifying variation driving the empirical results in the next section.

A full set of descriptive statistics for the shock variable, control variables, and the outcome variables appears in Appendix Table B1. The difference between the 90th percentile and the 10th percentile US employment shock is 7.5 percentage points, and the average municipio in our sample had an exposure to the US labor market (ξ_s) of approximately 25 percent, reflecting the fact that our sample uses municipios with relatively strong migrant ties to the US. We use these two facts below to help interpret the magnitudes of the estimated coefficients on US Employment shocks.

5 Results for Population Changes and Remittances

Our empirical analysis begins by finding the effect of US employment shocks on the overall growth of a municipio's population, following the comparative static modeled in Section 3. We then examine the contribution of both emigration and return migration to the total population response. Next, we use the same empirical specification to document additional effects on the demographic composition of the municipio population and the likelihood that households received remittance income. Together, the substantial effects on these initial outcomes represent channels through which migrant networks transmitted US local economic shocks to Mexican sending communities.

in 2005, and their residents received nearly 765,000 out of the roughly 923,000 MCAS identity cards issued in 2006. Appendix C.8 examines the robustness of the results to this sample choice and finds generally similar results when using the 1,194 municipios with at least 100 MCAS cards in 2006, without additional restrictions on population or US exposure. The notable exception is population growth, which has a weaker and statistically insignificant relationship with the US employment shock in this alternative sample.

5.1 Effects on Population Size and Migration

5.1.1 Population and Migration Measures

Our dependent variable for population growth is the proportional change in population ages 15-64 over a five-year interval, as measured every five years in the Census and in the Inter-Censal Count (Conteo). Both of these surveys also include questions about respondents' location of residence 5 years prior to the survey, allowing us to identify return migrants as those living in Mexico during the survey period and who lived in the US 5 years earlier.²⁵ Given the timing of the surveys, we can identify return migrants who moved from the US to Mexico during three five-year spans: 1995-2000, 2000-2005, and 2005-2010. We then measure return migration's contribution to population growth as the number of working-age return migrants to a given municipio, divided by the community's population at the start of the period. Note that this measure is not a traditional return migration rate, as the denominator is the municipio population rather than the number of people born in the municipio who were living abroad.

Our emigration outcome is the contribution of emigration to population growth, measured as the number of working-age emigrants during a five-year interval divided by the municipio's workingage population at the start of the interval. Information on emigration is not available in the 2005 Conteo, so we have emigration measures only from the 2000 and 2010 Censuses, which ask whether a household member emigrated to the US during during the five years prior to the date of the survey—1995–2000 or 2005–2010, respectively. This question is asked of the approximately 10% of the population who received a long-form survey, and it captures instances where one or more family members move to the US while some of the household remains in Mexico. We are unable to observe whole-household emigration.

5.1.2 Results for Population Changes

Table 1 provides estimates of Equation (7) using population growth and migration outcomes. Recall from Section 3 that we divide all dependent variables by the municipio's exposure to the US labor

²⁵The count of return migrants does not include those who moved to the US and back within the five-year window.

market, i.e. the share of people born in municipio s who were living in the US before the Great Recession. This adjustment accounts for the fact that population growth in sources with more people living in the US is more affected by any changes in the relative attractiveness of living at home or abroad.

Columns (1)-(3) provide results for population growth. The coefficient on the US employment shock is consistently negative and statistically significant, meaning that municipios connected to US destinations with larger job losses experienced larger increases in local population.²⁶ Column (1) presents the results of a regression of population growth from 2005-2010 on the US employment shock from 2006-2010 and Mexican-state fixed effects. In column (2), we control for population growth over the prior five-year period, allowing for pre-existing differences in population growth among municipios facing different demand shocks (see further discussion below in Section 5.1.4). Column (3) includes additional controls for destination immigration policies and for source-level trade shocks and homicide rates. The set of controls in column (3) may be affected by the US employment shocks, in which case this specification would be over-controlling. Nonetheless, while including these controls reduces the size of the coefficient of interest somewhat, it remains statistically significantly distinguishable from zero (p < 0.05).

Interpreting the magnitude of the coefficient requires two additional pieces of information: a difference in shock size, and a value for exposure to the US. From Appendix Table B1, the typical municipio in our sample had approximately 25 percent of its population living in the US, and the 90-10 percentile difference in shock size was 0.075. Therefore, the estimate in column (3) implies that when comparing two municipios with average exposure and a substantial difference in shock size, the more affected municipio experienced 2.1 percentage points faster population growth ((-1.125)(0.25)(-0.075) = 0.021). A similar calculation can be implemented to compare predicted outcomes for the pair of municipios shown in Figure 2. Dolores Hidalgo and Jaral del Progreso both have exposure to the US of around 0.3 and have a difference in shock size of roughly 0.1, predicting 3.4 percentage points faster population growth in Jaral del Progreso ((-1.125)(0.3)(-0.1) = 0.034).

²⁶When we allow the sample to include municipios with few migrants in the US (low values of ξ_s), the population estimate has smaller magnitude and loses statistical significance. See Appendix C.8.

To facilitate interpretation of the coefficients of interest, we provide similar calculations in all tables reporting the effects of the US employment shock on outcomes. The row labeled "Implied shock impact" multiplies the coefficient on the US employment shock by -0.01875 = (0.25)(-0.075). We also report the mean of the dependent variable (without dividing by exposure) for the quartile of municipios with the smallest declines in US employment demand. As an example of how these two values can be combined to understand the magnitude of the estimates, column (3) implies that the most-affected municipios saw population growth that was 18 percent higher compared to the least affected municipios (0.021/0.114).

5.1.3 Results for Return Migration and Emigration

Figure 4 shows that, in the aggregate, the decline in US labor demand was accompanied by both an increase in return migration to Mexico and a decline in emigration to the US. Following substantial net migration to the US in the 1990s and early 2000s, during 2005-2010 emigration to the US fell by 32 percent and return migration to Mexico quadrupled.²⁷

Columns (4)-(9) of Table 1 provide the results of estimating Equation (7) using return migration and emigration as outcomes. Because these measures are scaled by the initial municipio population, they can be interpreted as the contribution of each migration flow to local population growth. The coefficients on the US shock have the expected sign for both outcomes: municipios exposed to larger US job losses saw substantially larger population growth from return migration among people living in the US (columns (4)–(6)) and substantially less emigration of the local population to the US from 2005-2010 (columns (7)–(9)). The coefficients on the US employment shock are relatively stable across specifications, and the magnitudes are similar (although oppositely signed) for both the return migration and emigration outcomes. This similarity suggests that both return migration due

²⁷The substantial increase in the early 2000s has been documented elsewhere, including in Card and Lewis (2007), with explanations including the poor economic performance of Mexico after the ratification of the North American Free Trade Agreement (NAFTA) in 1990 and the Mexican Peso crisis of 1991 (Chiquiar and Salcedo 2013, Monras 2020b, Fajardo et al. 2017). Other analysis of higher-frequency data also shows a substantial slowdown over this time period, with annual net arrivals of fewer than 200,000 migrants (Passel, Cohn and Gonzalez-Barrera 2012). The nationwide emigration numbers shown in this figure are from CONAPO, which does not allow for the calculation of municipio-specific migration rates.

to lost jobs and potential migrants choosing not to leave for the US while demand was weak were important drivers of population adjustment in Mexican sources.

The estimated impact on the total population in column (3) is larger than the sum of the estimated contribution of increased return migration (6) and decreased emigration (9), a discrepancy we investigate in Appendix C.10. US Employment Shocks are not related to internal migration within Mexico nor to aging in to or out of the sample. Instead, the shocks are related to a residual component of population growth. This residual could come from unmeasured return migration—residents who were previously in the US failing to list that as their prior location—or unmeasured emigration of whole households. It could also represent statistical noise or some other channel of population adjustment. We nevertheless interpret Table 1 as showing that migration choices led to relative increases in population growth in the municipios most affected by the US Great Recssion by roughly 1–2 percentage points, with the lower bound the combined effects on return migration and emigration and the upper bound the measured effect on overall population.

5.1.4 Pre-Shock Trends in Population Changes

Table 2 provides an additional set of results useful for interpreting the estimates in Table 1. It examines the relationship between changes in municipio outcomes *prior* to the Great Recession and the US employment shocks those municipios would later face *during* the Great Recession. Ideally, these pre-Recession trends would be unrelated to subsequent shocks, as the sudden appearance of a relationship between the outcomes and the shocks would provide strong support for a causal interpretation. A pre-existing relationship between the outcome variables and a future shock does not necessarily indicate that an observed post-shock relationship is spurious, however, and controlling for the prior trend in the dependent variable avoids misinterpreting the simple continuation of a prior trend as the response to a shock.

Table 2 shows that pre-existing trends are more prominent for some outcomes than for others. Columns (1) and (2) indicate that municipios that would later experience larger declines in US employment already had somewhat smaller population growth during the pre-Recession period, which is consistent with the fact that specifications in Table 1 that include the pre-shock population growth control—columns (2) and (3)—have meaningfully smaller coefficient estimates on the US employment shock. In contrast, columns (3)-(6) find no such pre-existing relationship for return migration or emigration, again consistent with the findings in Table 1.

Because failing to account for pre-existing trends would present a challenge to the interpretation of the results, all of the subsequent sets of results include controls for the prior change in the dependent variable. These specifications therefore examine whether the connected labor market shocks led to a change in trend rather than simply asking whether the shocks are related to trends in the outcome variables around the the time of the shocks. For completeness, we provide pre-trend analyses analogous to Table 2 for all further outcomes in Appendix C.7.

5.2 Effects on Population Composition and Remittance Receipt

Along with effects on the *size* of local populations in Mexico, US employment declines may have altered the *composition* of the population if return migrants and discouraged emigrants had different characteristics than the overall population. Return migrants are identified in the Census, and we expect deferred emigrants, who are not identifiable, to have similar demographics. Appendix Section C.9 provides a descriptive comparison of return migrants to non-migrants, demonstrating that return migrants are much more likely to be male (69 percent vs. 49 percent) and are more likely to have primary-school education rather than higher or lower levels. They are also more likely to be married, and they have higher levels of labor force attachment, each of which is likely related to the fact that return migrants disproportionately fall in the 25–45 age range.

The first five columns of Table 3 examine the relationship between changes in the composition of source communities and the US employment shock. The positive coefficient estimate in column (1) implies that the sex ratio in a municipio facing the 90th percentile shock fell by 0.009 more than in a municipio at the 10th percentile, which is 23 percent of a standard deviation in the change in sex ratio over this time period.²⁸ The results in columns (2)-(5) of Table 3, however, show

²⁸Table 3 shows specifications with the full set of controls; alternative specifications appear in Appendix C.

no statistically significant relationship between the shock and the share of population with any particular level of education, despite the differing education levels of migrants and non-migrants.²⁹ Together, these results imply only a limited scope for the US shocks to affect wages in Mexican municipios because they primarily alter the aggregate amount of labor in a given municipio rather than the relative supplies of different skill levels. Consistent with this interpretation, we find no substantial wage effects in the next section.

Declines in US labor demand likely also decrease migrants' ability to send money back to Mexico. Column (6) of Table 3 examines the relationship between US shocks and the share of households receiving remittances from abroad in 2010. The positive and strongly statistically significant coefficient implies that households in the municipios facing larger declines in US labor demand were less likely to receive remittances, even after controlling for the baseline remittance share in 2000.³⁰ The point estimate of 0.47 in column (6) implies that, for municipios with average exposure to the US, a strongly affected community saw a roughly 1 percentage point larger decline in the share of households receiving remittances compared to a less affected community, which is a substantial decrease compared to the 5 percent mean among less-affected municipios.

Together, the results in this section show that US local labor demand shocks during the Great Recession affected Mexican sending communities through return migration, emigration, and remittance channels. In the following section, we examine how these changes in the size of the local labor force and the reduction in household budgets due to declining remittances affected employment, earnings, and household investment.

 $^{^{29}}$ These educational composition estimates are sensitive to controlling for the pre-Recession outcome measure (Appendix Table C3), so they should be interpreted with caution.

 $^{^{30}\}mathrm{The}\ 2005$ Conteo does not include questions regarding remittance receipt.

6 Labor Market and Investment Outcomes

6.1 Labor Market Outcomes

We examine impacts on local labor markets using full-count tabulations from the 2004 and 2009 Mexican Economic Census, which covers all formal economic activity in Mexico outside agriculture, livestock, forestry and a few service industries. Outcomes include municipio-level employment (separately by gender), aggregate yearly earnings, and aggregate yearly hours worked in covered sectors.³¹ We present municipio-level results here and provide results at the Mexican commuting zone level in Appendix C.6. Results are very similar regardless of the level of aggregation.

The loss of access to higher-wage jobs in the US is likely to affect local labor market outcomes in Mexico in three ways. First, a larger local working-age population increases both local labor supply and local labor demand, and the combination of these changes affects equilibrium wages and employment. Second, changes in net migration lead to important compositional shifts in the population of the municipio, with the average labor market attachment likely rising (Appendix Table C29). Finally, the loss of remittance income may lead some households to substitute into paid employment and away from home production.³²

Table 4 examines the net effect of these forces and finds substantial increases in employment and hours worked but minimal changes in hourly earnings. Column (1) of Table 4 examines the change in the municipio employment-to-population ratio from 2004 to 2009, using employment from the Economic Census and population from the 2005 Inter-Censal Count or 2010 Census, respectively. The negative coefficient estimates for the US employment shock imply that sources facing larger US employment declines exhibited larger increases in the employment to population ratio. Panels B and C make clear that the overall effect in Panel A is driven almost entirely by women.³³ The

³¹Service sectors that are not covered by the Economic Census include mass transit, taxis, farmers' insurance funds, political organizations, and domestic employees (INEGI 2009).

 $^{^{32}}$ It is also possible that these factors change workers' formality, but given the nature of our data, we are unable to examine this channel of adjustment empirically.

³³Although this analysis is not limited to married women, this result is similar to the "added worker effect" in which married women enter the labor force after their husbands lose employment. See Stephens (2002) for a thorough review of this literature. In Section 6.2, we present descriptive evidence suggesting that differences in labor supply are driven primarily by women in households with US migrants.

coefficient of -0.563 in Panel B implies that a strongly affected municipio with average exposure to the US experienced a 1.1 percentage point larger increase in employment to population ratio among women compared to a similar municipio that was less affected. Employment rates for men, however, did not change differentially based on the municipio's US labor demand shock, which suggests that the (largely male) return migrants and non-emigrants did not substantially crowd out employment in source communities. The reduction in household income from losing access to US jobs, however, likely led more women to enter the workforce.³⁴

Columns (2)-(4) demonstrate that local labor markets were able to accommodate substantial increases in the supply of hours worked without substantially reducing wages. The second and third columns show that both total municipio-level hours worked and total earnings increased in the most affected source communities, reflecting in part the increase in population shown in the previous section. Comparing two municipios at the 90th and 10th percentiles of the shock distribution, the more affected municipio experienced a 4.7 percentage point larger increase in local hours and a 5.2 percentage point larger gain in total earnings, both of which are meaningful changes compared to the average changes in less-affected municipios. Similar percentage effects on earnings and hours suggest minimal effects on hourly wages, which we confirm in the final column. The negative coefficient on the US shock implies a very small (and not statistically different from zero) increase in the average hourly wage rate for municipios facing more negative US shocks.³⁵

This set of results is somewhat surprising, as one may have expected the relative increase in local labor supply to negatively affect wages. To understand this result, we first note that the lack of wage impacts is not driven by compositional effects—Appendix Table C2 shows that the results in Panel A of Table 4 are qualitatively unchanged when including controls for changes in each municipio's demographic and educational composition.³⁶ Second, positively correlated labor

³⁴Although the focus of this paper is on economic adjustments immediately following the Great Recession, we are able to extend the analysis for this particular outcome through 2019. Appendix C.11 shows that both the increase in female employment and the lack of a change among men lasted through the end of that follow-up period.

³⁵With unweighted regressions, the estimate in column (4) of Table 4 is exactly the difference between the estimates in columns (2) and (3) – see Appendix Table C11.

³⁶The earnings data do not contain information about the characteristics of the workers, but we construct controls for changes in the gender mix, the age distribution (flexible bins), educational attainment (degree categories), and the local industry structure using Census data.

demand declines in migrant sources and destinations (not captured by our controls) are unlikely to explain the lack of negative wage effects, as this correlation would lead to negative bias on the US Employment Shock coefficient.

Instead, we interpret these results as consistent with the broader international migration literature, which typically finds modest effects of migration-related population growth on local equilibrium wages (National Academies of Sciences, Engineering, and Medicine 2017).³⁷ The implied relative increase in local population due to the loss of US jobs is roughly 2 percentage points for municipios whose shocks differ by the 90-10 percentile gap (see Table 1). This is a meaningful change but still substantially smaller than the 7 percent increase in local population in Miami due to the well-known Mariel Boatlift (Card 1990). Further, because there was no change in the skill mix in affected municipios (Table 3), the lack of a wage effect could be explained by modest capital adjustments over a five year period (Borjas 2013). Additionally, former migrants often return with lump-sum savings (Amuedo-Dorantes, Bansak and Pozo 2005), which could further stimulate local demand and mitigate downward wage pressure.

6.2 Supporting Evidence from Cross-Sectional Analysis

Our interpretation of the results in Table 4 presumes that the observed labor supply responses occurred in Mexican households with US migrants, as these were directly affected by a loss of US employment opportunities during the Great Recession. An ideal analysis would assess this interpretation using panel data to observe household-level *changes* in labor supply in response to return migration or a loss of remittances. Because no such panel dataset with municipio-level geography is available, we instead examine whether employment probabilities for members of households that had someone in the US during 2005-2010 are more strongly related to the US shocks compared to the same relationship among households without US migrants over that same time period.

We implement this complementary analysis using data from the 2010 Census. We define house-

³⁷Note that we are unable to disaggregate earnings or hours by gender or migration status, so we cannot estimate wage effects on particular subgroups. Studies finding a negative effect of migration on incumbent populations typically do so for narrow subgroups of workers, such as in Borjas and Doran (2012).

holds that were directly exposed to the US labor market during the Great Recession as those with either i) a member who had lived in the US in 2005 but returned to Mexico by 2010 or ii) a former household member who moved to the US after 2005. In both cases, we can be sure the household had a member in the US during the Great Recession period. Individuals living in households meeting this definition are identified by the indicator function $1(exposed_j)$ and referred to as "members of exposed households." We then use the following individual-level regression to ask whether the cross-sectional relationship between non-migrants' labor supply and US shocks is driven primarily by these individuals with direct ties to the US labor market.

$$1(\text{employed}_{j}) = \beta_{1}1(\text{exposed}_{j}) \cdot \left[\sum_{d \neq s} \varphi_{sd} \sum_{i} \frac{Emp_{id}^{M}}{Emp_{d}^{M}} \frac{d \ Emp_{id}}{Emp_{id}} \right]$$

$$+ \beta_{2}1(\text{exposed}_{j}) + \beta_{3} \left[\sum_{d \neq s} \varphi_{sd} \sum_{i} \frac{Emp_{id}^{M}}{Emp_{d}^{M}} \frac{d \ Emp_{id}}{Emp_{id}} \right]$$

$$+ \phi_{e(s)} + \Lambda \Delta X_{s} + \Pi \sum_{d \neq s} \varphi_{sd} \Delta X_{d} + \epsilon_{j}$$

$$(8)$$

If network-connected US job losses (the terms in square brackets) increase employment probabilities more for members of migrant households, the estimate of β_1 will be negative.

The results in Table 5 confirm this expected pattern. Columns (1) and (2) estimate the specification in (8), and column (3) estimates a more general specification subsuming all municipio-level terms into municipio fixed effects. In Panel A, which shows results using all residents, the interaction term's coefficient is negative and significant in column (1) (no controls), and column (2) shows that this result is robust to the inclusion of the full battery of controls.³⁸ The interaction term is still negative but statistically insignificant in column (3) with municipio fixed effects. Moreover, just as in Table 4, these employment effects among exposed households are driven almost entirely by women, for whom we find a significant negative effect even in the very demanding specification in column (3) of Panel B, with municipio fixed effects.³⁹

³⁸Results are qualitatively similar to those in columns (1) and (2) of Table 5, though a bit less precise, when controlling for state \times exposure status fixed effects.

³⁹The coefficient magnitudes in Tables 4 and 5 are not directly comparable due to different data sources and research design, and because our measure of $1(exposed_i)$ will not capture all households that were exposed to the

Figure 5 shows a binscatter plot visualizing the variation identifying β_1 in column (1) of Table 5. The gray circles plot the employment share of working-age population for those in unexposed households and the black diamonds show employment shares for those in exposed households. For unexposed households, there is no relationship between the employment probability and the US employment shock; this is expected because unexposed households by construction were not directly affected by US employment declines. In contrast, there is a strong negative relationship for members of exposed households. Together, these cross-sectional results support the interpretation that when migrants' households lost income due to negative US employment shocks, other household members, particularly women, sought to compensate by entering the labor force.⁴⁰ We note, however, that in some some specifications, there are similar patterns present in data from 2000 (see Appendix Table C22), which suggests interpreting this set of supporting results with caution.

6.3 Investment Results

If households are unable to fully offset a loss of US labor market income, they may adjust on other consumption and investment margins. Table 6 shows the effects of US labor demand shocks on two sets of investment behaviors: ownership of household durables and human capital investment via school attendance. Each column provides the results of a separate regression, returning to the specification in (7). The first four columns consider the change from 2005 to 2010 in the share of households owning the relevant household durable, including personal computers, washing machines, refrigerators, and televisions. The coefficient on the US employment shock for televisions is positive and significant, suggesting that households in more negatively affected municipios may have slowed down their television purchases. However, this estimate is sensitive to controlling for pre-Recession outcome growth (Appendix Table C6), so we encourage caution in interpreting this result.

The final three columns use municipio-level school attendance rates among different age groups primary (age 6-12), early secondary (13-15), and late secondary (16-18)—as the dependent variables.

US market.

⁴⁰Because Mexican population-level tabulations do not include earnings or hours separately by gender, we are not able to examine the gender wage gap.

The coefficients on the US employment shock are uniformly positive, meaning that declining US labor demand was associated with decreases in school attendance at all three levels. The coefficient is precisely measured only for the elementary school age outcome, however, where it implies a 0.4 percentage point smaller growth rate in school enrollment for a municipio with average exposure connected to a very negative shock compared to one with similar exposure but a mild shock.

Together, these results imply that the loss of access to a strong US labor market slowed investment in affected communities. These results are consistent with other research showing that sending communities' access to higher-paying foreign jobs improves children's schooling outcomes, especially Dinkelman and Mariotti (2016). Notably, in their context, the effects were longer-lasting, as the schooling gains continued even for cohorts who were of primary schooling age after workers lost access to the foreign labor market. The results in Table 6 comport with Caballero's (2022) findings, also in the US-Mexico context, in which school enrollment decreased in municipios with stronger migration ties to US destinations that adopted deportation policies.⁴¹ More generally, these findings are important because these differences in schooling attendance across municipios could lead to persistent earnings inequalities among children who were at pivotal schooling ages during the Great Recession.

7 Conclusion

This paper documents the role of migrant networks in transmitting the effects of the US Great Recession across the border to Mexico. In municipios whose migrants faced larger US labor demand declines, return migration increased more, emigration decreased more, and household remittances fell by more than in municipios facing smaller shocks. These changes in the local labor force, along with the reductions in household budgets due to lost remittance income, linked Mexican local labor market outcomes to US local labor demand shocks. The female employment-to-population ratio increased by more in harder-hit regions, likely as a way to compensate for lost US earnings among

 $^{^{41}}$ Caballero (2022) also provides a model clarifying the key channels through which return migration or deferred emigration are likely to affect schooling investment.

migrants. School enrollment for children age 6-12 also increased more slowly in these areas.

These findings demonstrate the substantial influence of the US labor market on Mexican demographic and economic outcomes, likely with long-lasting consequences. While this paper studies changes in US labor demand driven by the Great Recession, one can expect to observe similar effects if a large portion of Mexican migrants were to lose access to the US labor market due to changes in immigration and enforcement policies. For example, a well enforced universal E-Verify program would largely cut off labor market access for unauthorized immigrants, including approximately 43 percent of Mexican-born residents of the US in 2019 (Gonzalez-Barrera and Krogstad 2019).

Along with these policy implications, our findings inform the broader literature on the effects of immigration on local labor markets. Specifically, we find that aggregate outflows from sending locations are strongly responsive to labor demand conditions in the subset of US destinations where previous migrants from that source had historically settled. This finding conflicts with a key assumption behind the instrument most commonly used to correct for the endogeneity of local immigrant inflows to local labor demand conditions.⁴² The instrument treats aggregate inflows from each source as exogenous and focuses instead on resolving the potential endogeneity of migrants' location choices within the destination country, conditional on choosing to migrate. The finding that aggregate inflows from a source are endogenous to network-weighted demand suggests that US destinations with more positive values of the instrument may have systematically stronger unobserved labor demand growth. Finally, these findings reinforce the conclusion that US-Mexico migration operates through a series of tight connections between specific sources and destinations. Thus, we expect that any local shocks on one side of the border are likely to affect outcomes in migrant-connected localities on the other side.

⁴²This type of instrument was first introduced by Altonji and Card (1991), based on results in Bartel (1989). Although some papers attempt to identify specific source-level shocks to predict aggregate inflows from each source (Llull 2018), most papers simply assume that the total inflow from each source is exogenous. Jaeger et al. (2019) provide a more complete overview of this literature and offer an independent critique of the instrument based on the dynamics of adjustment to previous waves of migration.

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Figures and Tables

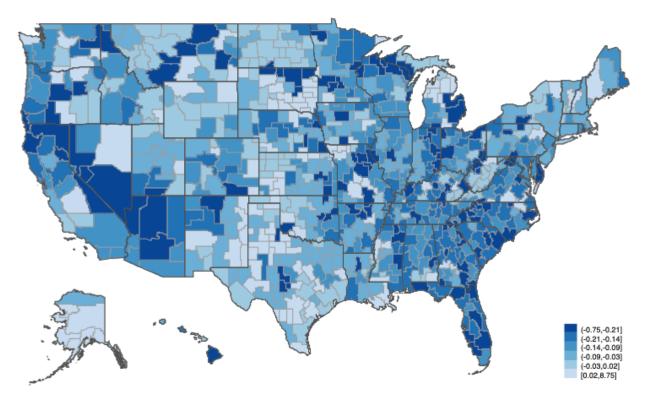


Figure 1: Change in US Employment across CZs

This map shows proportional changes in employment between 2006 and 2010 (bracketing the Great Recession) for each US commuting zone (CZ), with darker colors indicating larger declines in labor demand. Our measure accounts for Mexican-born workers' industry distribution of employment in each commuting zone (see main text). We use data from the County Business Patterns (CBP) supplemented with data from the American Community Survey (ACS) to fill in employment in a few industries that are not covered by the CBP. This variation in labor demand declines across US destinations leads to variation in network-connected labor demand across Mexican sources with different destination distributions (as shown in Figure 2).

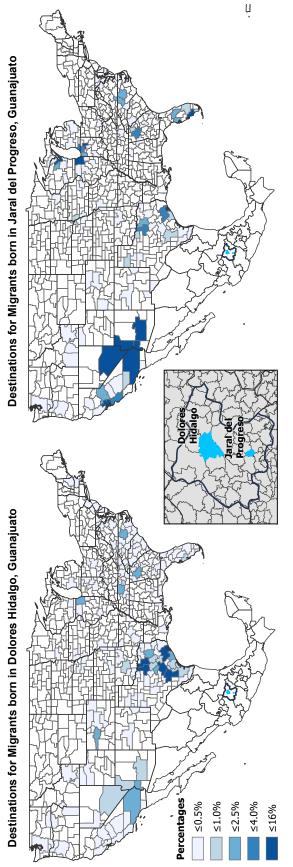
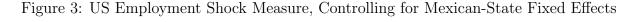
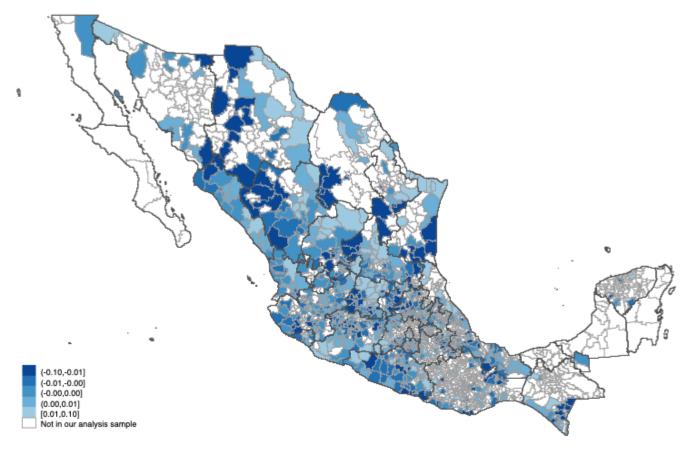




Figure 2: Example Migrant Destination Distributions for Two Municipios





This map shows the distribution of network-connected changes in US labor demand (as defined in the main text) over the time period of the Great Recession for each Mexican municipio, controlling for Mexican-state fixed effects. Our sample omits municipios (shown in white) with less than 5,000 residents in 2005, with initial exposure less than 0.066, or with fewer than 100 *matriculas* issued in 2006. This sample restriction maintains 56% of the year-2005 working age Mexican population. Because our analyses include Mexican-state fixed effects as control variables, the variation displayed in this map is the key identifying variation in our analysis.

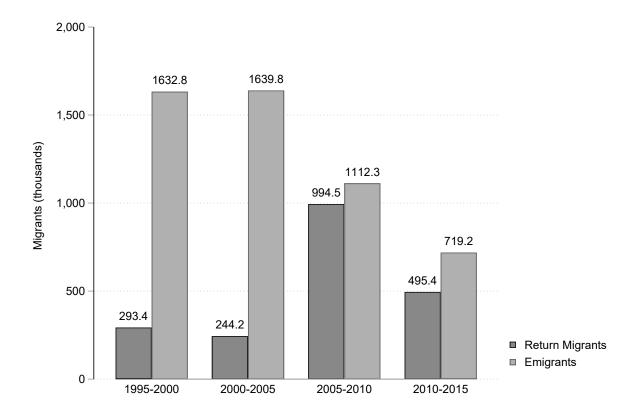


Figure 4: Five-Year Migration Flows Between Mexico and the US

Authors' calculations using data from INEGI and CONAPO. Return migration is calculated using the 2000 and 2010 Mexican Census, the 2005 Inter-Censal Count (*Conteo*), and the 2015 Inter-Censal Survey (*Encuesta Intercensal*). The return migration measure counts the number of people in Mexico reporting living abroad five years prior to the survey. Emigrants are calculated using data from the 2000 and 2010 Mexican Census and data from CONAPO for 2005 and 2015. The measure counts the number of emigrants leaving surveyed households within the five year period preceding the survey. Note the large increase in return migration to Mexico and the decrease in emigration from Mexico in 2005-10, during the US Great Recession.

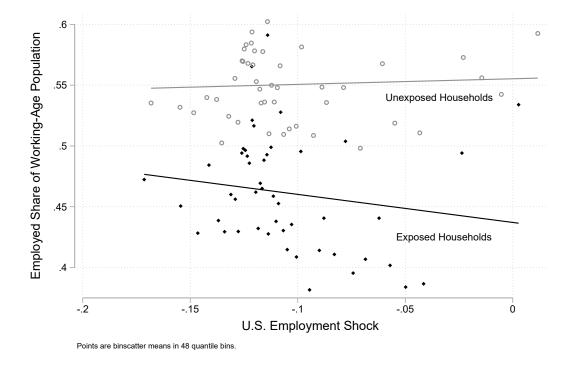


Figure 5: Employment Share of Working-Age Population vs US Employment Shock

This figure shows a binscatter plot of the variation identifying the main coefficient in column (1) of Table 5. Each gray circle shows the employment share of the working-age population living in households with no US migrants (unexposed households – see text for details) while each black diamond shows the employment share of the working-age population living in households with US migrants (exposed households). There is a strong negative relationship between the employment probability and the US employment shock for exposed households and essentially no relationship for unexposed households.

	Pop	oulation Gro	wth	Re	eturn Migra	tion		Emigratio	n
		2005-10	(2)	(1)	2005-10	(=)	()	2005-10	(*)
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
US Employment Shock	$\begin{array}{c} -2.070^{***} \\ [0.433] \\ (0.573) \end{array}$	$\begin{array}{c} -1.520^{***} \\ [0.454] \\ (0.527) \end{array}$	-1.224^{**} [0.482] (0.512)	-0.201** [0.095] (0.072)	-0.256*** [0.089] (0.066)	-0.171^{**} [0.090] (0.069)	0.210^{*} [0.132] (0.117)	$\begin{array}{c} 0.222^{*} \\ [0.121] \\ (0.116) \end{array}$	$\begin{array}{c} 0.287^{**} \\ [0.147] \\ (0.121) \end{array}$
Pre-shock Outcome		$\begin{array}{c} 0.683^{***} \\ (0.103) \end{array}$	$\begin{array}{c} 0.653^{***} \\ (0.104) \end{array}$		$\begin{array}{c} 1.476^{***} \\ (0.311) \end{array}$	$\begin{array}{c} 1.444^{***} \\ (0.308) \end{array}$		$\begin{array}{c} 0.124^{***} \\ (0.025) \end{array}$	$\begin{array}{c} 0.112^{***} \\ (0.025) \end{array}$
New 287g Policy			0.271^{*} (0.126)			$\begin{array}{c} 0.073^{***} \\ (0.017) \end{array}$			$0.025 \\ (0.025)$
Employment Policy			-0.057 (0.092)			-0.007 (0.011)			$0.002 \\ (0.017)$
Trade Shock			-0.079 (0.126)			0.023^{**} (0.010)			0.049^{***} (0.013)
Non-tradable share of Employment			-0.143^{**} (0.062)			-0.013* (0.007)			0.011 (0.012)
Homicide Rate 2005-10			-24.991^{**} (9.648)			-4.752^{***} (0.976)			-7.315^{***} (2.485)
Mean raw outcome	0.114	0.114	0.114	0.029	0.029	0.029	0.030	0.030	0.030
among less affected Implied shock impact State FE Observations R-squared	0.039 Yes 866 0.163	0.029 Yes 866 0.508	0.023 Yes 866 0.523	0.004 Yes 866 0.332	0.005 Yes 866 0.504	0.003 Yes 866 0.533	-0.004 Yes 866 0.271	-0.004 Yes 866 0.298	-0.005 Yes 866 0.320

Table 1: Population Growth, Return Migration, and Emigration

This table examines the effect of changes in US labor demand on the 2005-10 population growth, return migration to, and emigration from each Mexican source municipio. Note that outcome and pre-shock outcome variables are divided by exposure, ξ_s , as in equation (7). We restrict attention to individuals age 15-64. Population growth is defined as the proportional change in population. Return migration is the number of individuals reporting living in the US 5 years prior to the relevant survey, divided by the municipio population in the survey year, while emigration is the number of household members who left for the US during the 5 years prior to the relevant survey, divided by the initial municipio population, measured using the roughly 10% long-form sample from the 2000 or 2010 Census (emigration information is not available in 2005). We use full-count tabulations from the 2000 and 2010 Mexican Censuses and the 2005 Conteo. All specifications in columns (1) to (6) use a GLS re-weighting procedure to address potential heteroskedasticity. The "Pre-shock Outcome" controls in columns (2), (5), and (8) are 2000-2005 population growth, 2000-2005 return migration, and 1995-2000 emigration, respectively. Columns (3), (6), and (9) additionally control for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, and changes in homicide rates across municipios. All specifications control for Mexican state fixed effects. "Mean raw outcome among less affected" is the average of the dependent variable without dividing by exposure for municipios in the quartile with smallest magnitude US employment shocks. "Implied shock impact" provides the predicted difference in the outcome (without dividing by exposure) for municipios with the 90-10 percentile difference in shock size (0.075) and average exposure (0.25). Standard errors clustered at the Mexican commuting zone level are shown in parentheses. Due to the shift-share structure of the US Employment Shock, we also present Borusyak et al. (2022) standard errors for this variable in square brackets. *** p < 0.01, ** p < 0.05, * p < 0.1 based on standard errors in brackets when present.

	-	on Growth 00-05		Migration 0-05		gration 95-00
	(1)	(2)	(3)	(4)	(5)	(6)
US Employment Shock	-0.904**	-0.947**	0.023	0.044*	-0.072	0.063
	(0.426)	(0.479)	(0.024)	(0.025)	(0.142)	(0.160)
New 287g Policy		0.123		-0.003		0.038
		(0.133)		(0.007)		(0.051)
Employment Policy		-0.137*		0.008*		0.024
		(0.076)		(0.005)		(0.034)
Trade Shock		0.437***		0.007		-0.002
		(0.106)		(0.007)		(0.029)
Non-tradable share		-0.201***		-0.002		0.010
of Employment		(0.058)		(0.003)		(0.019)
Homicide Rate		-23.310***		-1.031		-9.909***
2005-10		(7.238)		(0.755)		(2.921)
State FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	866	866	866	866	866
R-squared	0.123	0.161	0.256	0.266	0.209	0.226

Table 2: Pre-trend Analysis: Population Growth, Return Migration, and Emigration

This table examines the effect of changes in US labor demand on the pre-shock population growth, return migration to, and emigration from each Mexican source *municipio* to determine whether there were pre-existing trends related to later shocks. Note that the pre-shock outcome variables are divided by exposure as in equation (7). We restrict attention to individuals age 15-64. Population growth is defined as the proportional change in population. Return migration is the number of individuals reporting living in the US 5 years prior to the relevant survey, divided by the municipio population in the survey year, while emigration is the number of household members who left for the US during the 5 years prior to the relevant survey, divided by the initial municipio population, measured using the roughly 10% long-form sample from the 2000 (emigration information is not available in 2005). We use full-count tabulations from the 2000 Census and the 1995 Conteo. All specifications in columns (1) to (6) use a GLS re-weighting procedure to address potential heteroskedasticity and control for Mexican state fixed effects. Standard errors clustered at the Mexican commuting zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

	\triangle Sex Ratio	\bigtriangleup Less than Primary	\triangle Primary	\triangle Secondary	\triangle University	Household
	(F/M) 2005-10	2005-10	2005-10	2005-10	2005-10	Remittances 2010
	(1)	(2)	(3)	(4)	(5)	(6)
US Employment Shock	0.479**	-0.019	0.049	-0.028	0.030	0.470***
o o Employment on our	[0.241]	[0.113]	[0.203]	[0.124]	[0.058]	[0.126]
	(0.200)	(0.115)	(0.168)	(0.076)	(0.062)	(0.125)
Pre-shock Outcome	-0.224***	0.541***	0.713***	0.838***	0.705***	0.447***
TIC-SHOCK Outcome	(0.032)	(0.025)	(0.067)	(0.037)	(0.035)	(0.045)
N 997 D I	0.025	0.000***	0.002**	0.017	0.001	0.002
New 287g Policy	-0.035 (0.048)	-0.082^{***} (0.029)	0.083^{**} (0.037)	-0.017 (0.018)	-0.001 (0.014)	0.003 (0.027)
	(0.048)	(0.029)	(0.037)	(0.018)	(0.014)	(0.027)
Employment Policy	0.010	0.009	-0.006	-0.009	0.006	0.022
	(0.037)	(0.021)	(0.029)	(0.012)	(0.011)	(0.018)
Trade Shock	-0.074***	-0.023*	0.290***	-0.180***	-0.017	0.044**
	(0.027)	(0.013)	(0.039)	(0.030)	(0.022)	(0.020)
Non-tradable share	0.015	0.002	0.080***	-0.008	-0.030***	0.023*
of Employment	(0.022)	(0.013)	(0.017)	(0.010)	(0.007)	(0.013)
Homicide Rate	10.482***	3.294**	5.139***	0.401	-1.032	-4.623*
2005-10	(3.664)	(1.641)	(1.977)	(0.866)	(0.966)	(2.478)
Mean raw outcome among less affected	-0.042	-0.053	0.003	0.038	0.018	0.051
Implied shock impact	-0.009	0.000	-0.001	0.001	-0.001	-0.009
State FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	866	866	866	866	866
R-squared	0.294	0.733	0.590	0.713	0.648	0.343

Table 3: Sex Ratio, Educational Attainment, and Households Receiving Remittances

This table examines the effect of changes in US labor demand on the 2005-10 change in the female to male sex ratio for the working age population (15-64), the 2005-10 change in the share of the working age population in each education level, and the share of households receiving remittances in 2010 for each Mexican source municipio. Note that outcome and pre-shock outcome variables are divided by exposure, ξ_s , as in equation (7). We measure the sex ratio and educational attainment using the 2000 or 2010 Mexican Census or 2005 Inter-Censal Count. We calculate the share of households receiving remittances as the number of households reporting receiving income from relatives living abroad divided by the municipio's total number of households in the Census year, using the 2000 or 2010 Mexican Census. All specifications in columns (1) to (6) use a GLS re-weighting procedure to address potential heteroskedasticity. The "Pre-shock Outcome" control in column (1) is the 2000-2005 change in the sex ratio. In columns (2)-(5) this control is the 2000-2005 change in the share of the municipio population with the listed level of schooling. In column (6), this control is the share of households receiving remittances in 2000. All specifications include controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, changes in homicide rates across municipios, and Mexican state fixed effects. "Mean raw outcome among less affected" is the average of the dependent variable without dividing by exposure for municipios in the quartile with smallest magnitude US employment shocks. "Implied shock impact" provides the predicted difference in the outcome (without dividing by exposure) for municipios with the 90-10 percentile difference in shock size (0.075) and average exposure (0.25). Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Due to the shift-share structure of the US Employment Shock, we also present Borusyak et al. (2022) standard errors for this variable in square brackets. *** p<0.01, ** p<0.05, * p<0.1 based on standard errors in brackets when present.

	A EDOD			
	\triangle EPOP	$\triangle \ln(\text{Hours})$ 2004-09	$\triangle \ln(\text{Earnings})$ 2004-09	$\triangle \ln(\text{EarnPerHour})$
	$\frac{2004-09}{(1)}$	(2)	(3)	2004-09 (4)
Panel A. All	(1)	(2)	(3)	(4)
US Employment Shock	-0.366**	-2.511*	-2.763	-0.266
05 Employment block	[0.160]	[1.459]	[2.330]	[1.816]
	(0.186)	(1.459)	(2.790)	(2.293)
	(0.100)	(1.100)	(2.150)	(2.230)
\triangle EPOP	-0.511***			
1999-04	(0.116)			
	()			
$\triangle \ln(\text{Hours})$		-0.159***	0.262**	0.435^{***}
1999-04		(0.046)	(0.122)	(0.100)
		· · · ·	· · · ·	
$\triangle \ln(\text{Earnings})$		0.071^{***}	-0.175***	-0.247***
1999-04		(0.023)	(0.062)	(0.049)
Mean raw outcome	0.013	0.126	0.345	0.219
among less affected				
Implied shock impact	0.007	0.047	0.052	0.005
<u>Panel B. Women</u> US Employment Shock	-0.563^{***} [0.137] (0.177)			
\triangle EPOP	-0.577***			
Women 1999-04	(0.159)			
Mean raw outcome	0.015			
among less affected				
Implied shock impact	0.011			
Panel C. Men US Employment Shock	-0.112 [0.250] (0.236)			
\triangle EPOP	-0.408***			
Men 1999-04	(0.087)			
Mean raw outcome	0.009			
among less affected				
Implied shock impact	0.002			
State FE	Yes	Yes	Yes	Yes
Observations	865	846	846	846

Table 4: Employment-to-population Ratio and Earnings per Hour

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio, earnings, hours worked, and earnings per hours in each municipio, using employment, earnings and hours from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. In Appendix Table C16, we implement the same analysis at the Mexican Commuting Zone level, which may better approximate local labor markets, finding similar results. Note that the outcome and pre-shock outcome variables are divided by exposure, ξ_s , as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (4) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, changes in homicide rates across municipios, and Mexican state fixed effects. "Mean raw outcome among less affected" is the average of the dependent variable without dividing by exposure for municipios in the quartile with smallest magnitude US employment shocks. "Implied shock impact" provides the predicted difference in the outcome (without dividing by exposure) for municipios with the 90-10 percentile difference in shock size (0.075) and average exposure (0.25). Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Due to the shift-share structure of the US Employment Shock, we also present Borusyak et al. (2022) standard errors for this variable in square brackets. *** p<0.01, ** p<0.05, * p<0.1 based on standard errors in brackets when present. 45

	State FE	State FE	Municipio FE
	(1)	(2)	(3)
<u>Panel A. All</u>			
US Employment $Shock^* \mathbb{1}(exposed_h)$	-0.189^{***}	-0.139**	-0.101
	(0.069)	(0.066)	(0.068)
$\mathbb{1}(exposed_h)$	-0.038***	-0.031***	-0.020**
	(0.008)	(0.008)	(0.008)
Observations	$33,\!270,\!660$	$33,\!270,\!660$	$33,\!270,\!660$
Panel B. Women			
US Employment $Shock^* \mathbb{1}(exposed_h)$	-0.297***	-0.224***	-0.201**
	(0.086)	(0.084)	(0.087)
<i>.</i>			
$\mathbb{1}(\operatorname{exposed}_h)$	-0.048***	-0.038***	-0.026***
	(0.010)	(0.010)	(0.010)
Observations	17,511,744	17,511,744	17,511,744
$\frac{Panel C. Men}{UC Employment Charles 1 (compared)}$	0.054	0.005	0.001
US Employment $Shock^* \mathbb{1}(exposed_h)$	-0.054	-0.025	0.021
	(0.087)	(0.085)	(0.082)
1(opposed)	-0.007	-0.003	0.005
$\mathbb{1}(\operatorname{exposed}_h)$			
	(0.009)	(0.009)	(0.009)
Observations	15,758,916	15,758,916	15,758,916
00501 vali0115	10,700,910	15,156,910	10,100,910

Table 5: Cross-Sectional Employment Analysis 2010

This table examines how labor supply behavior differs for households with and without US migrants in municipios facing different US shocks. We use cross-sectional data from the 2010 Census and define households exposed to US labor markets as those with either return migrants or with a household member living in the US. Columns (1) and (2) estimate the specification in Equation (8), including the main effect of the US Employment shock as a control, while column (3) estimates a more general specification with municipio fixed effects and thus omits the US Employment Shock main effect. Column (2) shows the results including controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, and changes in homicide rates across municipios (and the municipio fixed effects in column (3) subsume all these controls). Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

	\triangle Computer 2005-10 (1)	\triangle Washing Machine 2005-10 (2)	\triangle Refrigerator 2005-10 (3)	$\begin{array}{c} \bigtriangleup 1.V\\ 2005-10\\ (4) \end{array}$	\triangle Attendance Kate (ages 6-12) 2005-10 (5)	\triangle Attendance Rate (ages 13-15) 2005-10 (6)	\triangle Attendance Kate (ages 16-18) 2005-10 (7)
US Employment Shock	-0.122	0.057	0.211	0.474^{**}	0.200^{***}	0.371	0.011
	[0.096]	[0.246]	[0.248]	[0.241]	[0.082]	[0.296]	[0.348]
	(0.101)	(0.226)	(0.225)	(0.184)	(0.070)	(0.246)	(0.241)
\triangle Outcome	0.873^{***}	0.279^{***}	0.337^{***}	0.400^{***}	-0.016	0.114^{***}	0.176^{***}
2000-05	(0.022)	(0.021)	(0.022)	(0.027)	(0.024)	(0.028)	(0.027)
New 287g Policy	-0.030	0.064	0.052	0.012	0.025	0.086	0.035
	(0.056)	(0.057)	(0.048)	(0.017)	(0.058)	(0.058)	
Employment Policy	-0.036	-0.079**	-0.028	0.012	0.013	0.053	0.002
	(0.039)	(0.041)	(0.033)	(0.012)	(0.040)	(0.038)	
Trade Shock	-0.057	0.025	0.057	0.044	0.040^{***}	0.115^{***}	-0.010
	(0.056)	(0.041)	(0.031)	(0.013)	(0.027)	(0.037)	
Non-tradable share	-0.019	-0.045	-0.011	0.012	-0.008	-0.055**	-0.086***
of Employment	(0.029)	(0.037)	(0.030)	(0.008)	(0.026)	(0.030)	
Homicide Rate 05-10	0.512	1.418	5.125^{*}	-0.957	-7.930**	-3.250	
	(2.906)	(2.778)	(2.820)	(1.176)	(4.008)	(4.501)	
Mean raw outcome among less affected	0.065	0.061	0.060	0.024	0.006	0.056	0.060
Implied shock impact	0.002	-0.001	-0.004	-0.009	-0.004	-0.007	-0.000
State FE	Yes	Yes	Yes	\mathbf{Yes}	Yes	Yes	Yes
Observations	866	866	866	866	866	866	866
R-squared	0.848	0.580	0.546	0.500	0.096	0.305	0.330

Attendance
School
and
Ownership
Appliance (
Table 6:

This table examines the effect of changes in US labor demand on the 2005-10 change in ownership of household durables (personal computers, washing machines, retrigerators, and televisions) and in school attendance. We calculate the change in the share of households owning the relevant household durable and the share of the population in primary (age 6-12), secondary (age 13-15), and high-school (age 16-18) reporting having attended school using the 2000 or 2010 Mexican Census or 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure, ξ_s , as in equation (7). All specifications in columns (1) to (7) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, changes in homicide rates across municipios, and Mexican state fixed effects. "Mean raw outcome among less affected" is the average of the dependent variable without dividing by exposure for municipios in the quartile with smallest magnitude US employment shocks. Shock, we also present Borusyak et al. (2022) standard errors for this variable in square brackets. *** p<0.01, ** p<0.05, * p<0.1 based on standard errors in brackets when "Implied shock impact" provides the predicted difference in the outcome (without dividing by exposure) for municipios with the 90-10 percentile difference in shock size (0.075) and average exposure (0.25). Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Due to the shift-share structure of the US Employment present.

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(Not for publication)

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A Model Derivations

A.1 Derivation of Equation 3

To derive equation (3), start with the total derivative of $M_{ss} = M_s P_s(s)$, holding M_s constant.

$$dM_{ss} = M_s \left[\frac{\partial P_s(s)}{\partial v_s} dv_s + \sum_{d \neq s} \frac{\partial P_s(s)}{\partial v_d} dv_d \right]$$
(9)

Then, evaluate the partial derivatives of the choice probabilities in (2).

$$\frac{\partial P_s(s)}{\partial v_s} = \frac{\alpha \exp(\alpha v_s + n_{ss})}{\sum_{d'} \exp(\alpha v_{d'} + n_{sd'})} - \alpha \left(\frac{\exp(\alpha v_s + n_{ss})}{\sum_{d'} \exp(\alpha v_{d'} + n_{sd'})}\right)^2$$

$$= \alpha P_s(s)(1 - P_s(s))$$
(10)

$$\frac{\partial P_s(s)}{\partial v_d} = -\alpha \frac{\exp(\alpha v_s + n_{ss}) \exp(\alpha v_d + n_{sd})}{\left(\sum_{d'} \exp(\alpha v_{d'} + n_{sd'})\right)^2} \quad \text{where } d \neq s$$

$$= -\alpha P_s(s) P_s(d) \tag{11}$$

Plugging these into (10) and simplifying yields (3).

A.2 Derivation of Equation 6

To derive equation (6), start with its left hand side, and impose the assumptions listed just above (6). Under assumption i), the employment probability facing Mexican workers is the Mexican employment to population ratio, so

$$Pr(emp_d) = \frac{Emp_d^M}{M_d} = \frac{\sum_i Emp_{id}^M}{M_d},$$
(12)

where Emp_{id}^{M} is Mexican employment in industry *i* in destination *d* and M_d is the Mexican-born population of *d*. Take the derivative of $w_d Pr(emp_d)$, holding wages fixed under the rigid wage assumption.

$$w_d dPr(emp_d) = w_d \frac{\sum_i dEmp_{id}^M}{M_d} = \sum_i w_d \frac{Emp_d^M}{M_d} \frac{Emp_{id}^M}{Emp_d^M} \frac{dEmp_{id}^M}{Emp_{id}^M} = \sum_i w_d \gamma_d \frac{Emp_{id}^M}{Emp_d^M} \frac{dEmp_{id}^M}{Emp_{id}^M}, \quad (13)$$

where $\gamma_d \equiv Emp_d^M/M_d$ is the employment probability among Mexican-born workers in d. Note that $w_d\gamma_d$ is the expected earnings among Mexicans in d. Assumption iii) implies that job losses in a given industry and location are allocated proportionately to Mexican-born and other workers, so

$$\frac{dEmp_{id}^M}{Emp_{id}^M} = \frac{dEmp_{id}}{Emp_{id}}.$$
(14)

Assumption ii) implies equal baseline expected earnings across destinations, so

$$w_d \gamma_d \equiv \delta. \tag{15}$$

Plugging the preceding two expressions into (13) yields the right hand side of (6).

A.3 Alternative Interpretation of Shock Magnitude

As mentioned in Section 3.2, the US employment shock measure in (6) can be interpreted as the wage times the number of US jobs lost per migrant under the assumption that the wage is constant across US locations, i.e. $w_d = w$, $\forall d$, which in turn implies that $\gamma_d = \gamma$, $\forall d$ and $\delta = w\gamma$. Given this assumption, start with the right of (6) and use the definition of γ from above, the migration network term $\varphi_{sd} = M_{sd}/(\sum_{d'\neq s} M_{sd'})$, and the the identity

$$\frac{d \ Emp_d^M}{Emp_d^M} \equiv \sum_i \frac{Emp_{id}^M}{Emp_d^M} \frac{d \ Emp_{id}^M}{Emp_{id}^M}.$$
(16)

Plug these in and simplify to yield

$$w\gamma \sum_{d \neq s} \varphi_{sd} \sum_{i} \frac{Emp_{id}^{M}}{Emp_{d}^{M}} \cdot \frac{d \ Emp_{id}}{Emp_{id}} = w \sum_{d \neq s} \frac{Emp_{d}^{M}}{M_{d}} \frac{M_{sd}}{(\sum_{d' \neq s} M_{sd'})} \frac{d \ Emp_{d}^{M}}{Emp_{d}^{M}}$$

$$= w \frac{1}{\sum_{d' \neq s} M_{sd'}} \sum_{d \neq s} \frac{M_{sd}}{M_{d}} d \ Emp_{d}^{M}.$$
(17)

Since $\sum_{d'\neq s} M_{sd'}$ is the total number of US migrants from s, the right side of (17) is the wage times the number of jobs lost (or gained) per migrant from s. This derivation thus shows that, under the appropriate assumptions, the US employment shock facing each Mexican municipio has an intuitive reduced-form interpretation as the wage times the number of US jobs lost per migrant.

B Data

B.1 Matrículas Consulares de Alta Seguridad

We use a custom extract from the MCAS administrative dataset covering all the *matrículas con*sulares issued in 2006 to Mexican-born individuals by place of birth in Mexico and place of residency in the US. Because this extract did not contain numerical identifiers for municipio or county, we needed to determine which municipio each record represented. We assigned each source municipio name an identifier to match those used by Mexico's Statistical Office (INEGI) and we assigned each US count a county-level FIPS code.

Mexican place of birth: The extract contained Mexican state and municipio of birth. However, the field for municipio of birth was sometimes reported by cardholders as their town or place of birth. After merging municipio names from INEGI's list, we ended up with 87 percent of the matrículas perfectly matched. For the remaining 13 percent of the matrículas, more than half (7 percent of the total) were from individuals reporting Mexico City as their place of birth. To address this issue, we aggregated *municipios* within Mexico City (Distrito Federal) in all Mexican datasets. For the remaining 6 percent we matched the information recorded in the municipio field to INEGI's identifiers by using a record linkage method (reclink2) in Stata, performing fuzzy matches. With these two procedures we were able to identify 95 percent of *municipios* in the dataset. Finally, we manually assigned places to *municipios* for 3 percent of the unmatched matrículas in the data, leaving us with 98 percent of the matrículas matched to a *municipio*.

US county of residency: The extract contained US state and county of residency. However, in some instances cardholders reported places or cities of residency, abbreviated or misspelled city names (i.e. LA for Los Angeles), or in very few instances county of residency that did not corresponded to the reported state of residence (i.e. Charleston, South Dakota instead of Charleston, South Carolina). After merging county names using the FIP codes list from the US Census Bureau, we ended up with 88 percent of the matrículas perfectly matched. For the remaining unmatched cases, including those just mentioned, we manually coded the correct counties.

We aggregate destination counties to the commuting zone level, using the crosswalk in Dorn (2009).⁴³ This provides us with information on the connections between each Mexican municipio and each US commuting zone.

B.2 US Employment

We measure changes in US labor demand using payroll employment from the County Business Patterns (CBP) data from 2006-2010. This dataset includes the universe of employment at business establishments in covered industries in each US county and is thus the most accurate data source for measuring local employment declines in the private sector. Unfortunately, the CBP data does not include any demographic information and we therefore cannot use it to directly calculate job losses among Mexican-born workers. We therefore combine the CBP data with data from the 2006 American Community Survey (ACS) from IPUMS (Ruggles, Flood, Goeken, Grover, Meyer, Pacas and Sobek 2017). The ACS data allow us to fill in employment changes for industries not covered by the CBP (including government and agriculture) and to measure the CZ-specific share of Mexicanborn individuals working in each industry. We observe 20 separately identifiable industries at the

 $^{^{43}}$ https://www.ddorn.net/data.htm (file E7)

CZ level. Because our goal is to measure labor demand changes specific to Mexican-born workers employed in US industries, we also use data from the 2006 American Community Survey (ACS) to measure the share of the Mexican-born migrants employed in each industry prior to the Great Recession.

Because we want to measure changes in US labor demand at the local labor market level, we use commuting zones as our geographic unit of analysis in the US. We aggregate county-level employment information from CBP and migrant destination data from MCAS to the commuting zone level using the crosswalk in Dorn (2009).⁴⁴ We make manual adjustments to maintain consistent county boundaries over time. Because the most disaggregated sample available in the ACS is the Public Use Micro Areas (PUMA), we use another crosswalk from Dorn (2009) to match PUMAS to commuting zones.⁴⁵ After these aggregations at the commuting zone level, we were able to use these two datasets to construct a weighted average of employment changes accounting for the industrial composition of Mexican employment in each US labor market.

B.3 Demographic and Population Outcomes

Return Migration: We define return migrants from the US to each Mexican municipio between 2005-2010 as individuals ages 15-64 living in Mexico during the 2010 Census reference period but who lived in the US five years before. These flows are identified through a question that asks respondents their country of residency five years prior to the Census year. Note that the count of return migrants does not include any individuals who were living in Mexico five years previously but who moved to the US and back within the five year window. We also calculate pre-shock return migration for the working age population between 2000-2005 using information from the the 2005 *Conteo* and the 2000 Mexican Census. To calculate these measures, we use official tabulations of the full-count 2010 Mexican Census and the 2005 Inter-Censal Count (*Conteo*), available at INEGI's website. We divided the 2005-2010 measure by the 2005 working age population and the 2000-2005 measure by the 2000 Mexican Census and 2005 *Conteo*, available at INEGI's website.

Emigration: We calculate emigration from each Mexican municipio to the US as the number of individuals ages 15-64 who reported leaving between 2005-2010 in the 2010 Mexican Census, using the 2010 Mexican Census supplemental sample questionnaire. This survey, available at the Mexican Statistical Office website (INEGI) contains an international migration module, conducted on a 10% sample of Mexican households in each Census year, asking respondents if anyone in the household went to live in the US during the previous five years. We also calculate pre-shock emigration for the working age population who reported leaving to the US between 1995-2000 using microdata from the 2000 Mexican Census, as the 2005 *Conteo* does not include information on emigration. We divided the 2005-2010 measure by the 2005 working age population and the 1995-2000 measure by the 2000 working age population in each municipio, using official tabulations of the full-count 2000 Mexican Census and 2005 *Conteo*, available at INEGI's website. Note that because the Mexican Census does not provide information on emigration of entire households, our emigration measure may be underestimated. Since we construct the emigration measure using a sample rather than the population, the return migration estimate is likely more reliable.

⁴⁴https://www.ddorn.net/data.htm (file E7)

⁴⁵https://www.ddorn.net/data.htm (file E5).

Population Changes: This measure is defined as the proportional change in the total number of Mexican individuals ages 15-64 in each Mexican municipio between 2005-2010 and between 2000-2005. We use official tabulations of the full-count 2000 and 2010 Mexican Census and the 2005 Conteo, available at INEGI's website.

Sex Ratio: This measure is defined as the ratio of the total number of Mexican women ages 15-64 in each Mexican municipio at the time of the Census or *Conteo* to the total number of men ages 15-64 in each Mexican municipio at the same time. We calculate the female to male sex ratio using official tabulations of the full-count 2000 and 2010 Mexican Census and the 2005 *Conteo*, available at INEGI's website.

Educational Attainment: We measure the share of the municipio population with each education level as the total number of individuals ages 15-64 with that level of education at the time of the Census or *Conteo*, divided by the total working age population of the same municipio. The "less than primary education" category includes those with no schooling and with up to 4 years of primary education; individuals with primary education are those with primary and lower secondary completed; those with upper secondary completed are considered to have a secondary education; individuals with university education are those with at least some post-secondary schooling. We use official tabulations of the full-count 2000 and 2010 Mexican Census and the 2005 *Conteo*, available at INEGI's website.

Households receiving Remittances: We measure the share of households receiving remittances as the proportion of households in each municipio and Census year reporting receiving income from relatives abroad. We use 2000 and 2010 Mexican Census microdata from (Ruggles et al. 2017) to calculate this measure. Note that although the 2000 Census includes a question on the amount of remittances received by each surveyed household in Mexico, the 2010 Census reports only whether the household received any remittances. The relevant question is somewhat open-ended regarding the timeframe of remittance receipt, and enumerators encouraged respondents to report both regular and sporadic remittance receipt.

B.4 Economic Outcomes

Employment-to-population ratio: We calculate the employment-to-population ratio for each municipio as the share of the population ages 15-64 with formal employment. We measure the number of employed people in each municipio, separately for women and men, using full-count tabulations, available at INEGI's website, from the 1999, 2004, and 2009 Mexican Economic Census, which covers formal employment in Mexico excluding agriculture, livestock, forestry, mass transit, taxis, farmers' insurance funds, political organizations, and domestic employees. We divide this employment count by the working age population of the corresponding municipio using official tabulations of the full-count 2000 and 2010 Mexican Census and the 2005 *Conteo* available at INEGI's website.

Earnings per hour: We measure earnings per hour as the municipio's aggregate yearly earnings divided by the municipio's aggregate yearly hours worked, using full-count tabulations from the 1999, 2004, and 2009 Mexican Economic Census. This earnings measure therefore covers the same sectors as the employment measure. Note that earnings and hours worked are not available separately for men and women.

Appliance Ownership: We calculate appliance ownership as the proportion of households in each municipio reporting owning the relevant appliance at the time of the Census or *Conteo*, including a personal computer, a refrigerator, a washing machine, or a television. We use official tabulations of

the full-count 2000 and 2010 Mexican Census and the 2005 *Conteo*, available at INEGI's website.

School Attendance Rate: We calculate the school attendance rates as the total number of children in each municipio attending primary (ages 6-12), lower secondary (ages 13-15) or upper secondary (ages 16-18) education at the time of the Census or *Conteo*, divided by the total population in that same age group and municipio. We use official tabulations of the full-count 2000 and 2010 Mexican Census and the 2005 *Conteo*, available at INEGI's website.

B.5 Controls

Immigration policies: We use indicators for newly introduced state-level anti-immigrant employment legislation and indicators for new 287(g) agreements allowing local officials to enforce federal immigration law, using information from the database complied by Bohn and Santillano (2017). Because these variables are measured at the state level, we use the crosswalk in (Dorn 2009) to map states to commuting zones.⁴⁶ The control variables we include in the regressions are weighted averages of changes in the policy indicators with weights based on the destination distribution of migrants from the relevant municipio.

Trade Shocks: We focus on the effects of declining US employment opportunities facing potential migrants from Mexico, but the Great Recession also reduced trade between Mexico and the US Because we focus on migration-related channels, the effects of declining trade could confound our analysis. We control for such trade effects by constructing municipio-level exposure to change in Mexican trade with the US We begin by constructing industry-level changes in trade from Mexico to the US per Mexican worker. We use trade data from the US Census Bureau, provided in Stata format by Peter Schott.⁴⁷ We aggregate these data from 10-digit HS products to 4-digit NAICS industries using the concordance from Pierce and Schott (2012) and calculate the change in trade value from the period 2001-2005 to the period 2006-2010. We measure initial Mexican employment using data from the 2004 Mexican Economic Census, which covers the vast majority of firms in sectors outside agriculture.⁴⁸ For each municipio, we then generate a weighted average of these industry-level trade changes, where the weights reflect the municipio's 2004 distribution of tradable-industry employment across 4-digit NAICS industries, also calculated using the Economic Census. The weights sum to one across tradable industries, and we include an additional control for the nontradable share of employment in 2004 to address the incomplete shares problem (Borusyak et al. 2022).

Homicide Rate: We measure the number of homicides during 2005-2010 for each municipio divided by the 2005 population for the corresponding municipio, by using administrative yearly records from the Mexican Statistical Office (INEGI) and official tabulations of the full-count 2005 *Conteo* available at INEGI's website.

⁴⁷https://sompks4.github.io/sub_data.html

⁴⁶https://www.ddorn.net/data.htm (file E8)

⁴⁸https://www.inegi.org.mx/app/saic/ Accessed March 27, 2020.

B.6 Summary Statistics

Descriptive statistics for the shock measure and for control variables appear in Panel A of Appendix Table B1. Panel B displays summary statistics for outcome variables. Note that the statistics listed in Panel B are calculated using the outcome variables prior to dividing by exposure.

	Obs	Mean	Std. Dev.	p10	<u></u>
Panel A: Shock Measure and Control Variables	Obs	mean	Stu. Dev.	pro	p90
Exposure	866	0.259	0.138	0.096	0.453
US Employment Shock	866	-0.108	0.034	-0.139	-0.064
New 287g Policy	866	0.133	0.128	0.032	0.254
Employment Policy	866	0.133 0.174	0.120 0.137	0.002 0.047	0.360
Trade Shock (\$1000s)	866	-8.479	70.479	-13.747	4.706
Non-tradable share of Employment	866	0.722	0.175	0.467	0.902
\triangle Homicide Rate (per 1000)	866	0.891	1.329	0.055	2.174
Panel B: Outcome Variables					
Population growth and Migration - Mexican	Popula	tion Cen	isus		
Population Growth 2005-10	866	0.138	0.079	0.065	0.218
Return Migration 2005-10/Pop2005	866	0.028	0.017	0.008	0.051
Emigration $2005-10/Pop2005$	866	0.024	0.021	0.006	0.047
Other Channels - Mexican Population Censu	s				
\triangle Sex Ratio (F/M) 2005-10	866	-0.046	0.040	-0.098	-0.002
\triangle Less than primary education 2005-10	866	-0.053	0.020	-0.079	-0.025
\triangle Primary education 2005-10	866	0.004	0.040	-0.049	0.051
\triangle Secondary education 2005-10	866	0.035	0.014	0.020	0.052
\triangle University education 2005-10	866	0.020	0.012	0.006	0.036
Households with Remittances 2010	866	0.040	0.036	0.008	0.093
Economic Outcomes - Mexican Economic Ce	nsus				
\triangle Log Earnings per Hour 2004-09	864	0.230	0.477	-0.243	0.659
\triangle Epop 2004-09	866	0.018	0.049	-0.012	0.058
\triangle Epop Men 2004-09	866	0.014	0.054	-0.030	0.061
\triangle Epop Women 2004-09	866	0.021	0.054	-0.003	0.060
Economic Outcomes - Mexican Population C	ensus				
\triangle Computer Ownership 2005-10	866	0.066	0.034	0.026	0.113
\triangle Washing Machine Ownership 2005-10	866	0.068	0.043	0.012	0.123
\triangle Refrigerator Ownership 2005-10	866	0.064	0.045	0.008	0.127
\bigtriangleup TV Ownership 2005-10	866	0.027	0.039	-0.003	0.077
\bigtriangleup Attendance Rate (ages 6-12) 2005-10	866	0.005	0.011	-0.007	0.018
\bigtriangleup Attendance Rate (ages 13-15) 2005-10	866	0.050	0.040	0.003	0.106
\triangle Attendance Rate (ages 16-18) 2005-10	866	0.055	0.037	0.008	0.102

Table B1: Summary Statistics

This table shows summary statistics for all municipio-level outcomes, the main independent variable, and control variables used in the analyses. To construct dependent variables used in the regressions, we divide the listed variables by exposure. The table presents outcome values prior to dividing by exposure.

C Additional empirical results

C.1 Shock variation across CZs

In Table C1 we present summary statistics describing the sources of variation in the employment shocks across US commuting zones (CZs). Recall from section 2.1 that we measure US employment changes facing Mexican-born workers in commuting zone d as

$$\sum_{i} \frac{Emp_{id}^{M}}{Emp_{d}^{M}} \left(\frac{Emp_{id}^{2010} - Emp_{id}^{2006}}{Emp_{id}^{2006}} \right), \tag{18}$$

where Emp_{id}^{year} is employment in industry *i* in destination commuting zone *d* and Emp_{id}^M/Emp_d^M is the share of Mexican-born workers in commuting zone *d* working in industry *i* in 2006.

Because this measure weights each industry based on its share of Mexican-born employment, Emp_{id}^M/Emp_d^M , industries with a larger share of Mexican-born employment have more influence on the shock measure. The first column of Table C1 lists each industry's share of Mexican-born employment at the national level, showing that Construction, Hotel and Dining, and Manufacturing account for the largest shares of Mexican employment in the US. The second column then reports the variance in each industry's employment share across US CZs. In this case, Agriculture, Construction, and Manufacturing are the top three sectors, indicating that these industries are more concentrated in a particular set of CZs than are other industries. Finally, the third column reports the cross-CZ variation in each industry's local employment growth from 2006 through 2010 (the term in parentheses in equation (18)). By far the largest entry is for manufacturing, indicating wide variation in local manufacturing employment growth across CZs. Administration and Education exhibit the next largest variance in local employment growth figures.

The information in Table C1 shows that there was substantial variation in employment growth within industries across locations, particularly in manufacturing, and that the employment mix of Mexican-born workers also differed across CZs. These two sources of variation combine to generate cross-CZ variation in the employment shock facing Mexican-born residents of each US location.

Industry	Employment Share Mexican-born	Variance in Employment Share Mexican-born	Variance in Industry Employment Growth
Administration	8.826	0.080	0.097
Agriculture	5.736	0.220	0.080
Arts, Recreation	1.186	0.031	0.029
Construction	21.473	0.205	0.065
Education	2.580	0.134	0.093
Finance	1.185	0.015	0.002
Government	0.846	0.058	0.014
Health Care	4.454	0.141	0.044
Hotel, Dining	13.104	0.158	0.022
Information	0.631	0.042	0.009
Management	0.038	0.010	0.002
Manufacturing	16.701	0.247	0.387
Mining	0.312	0.033	0.022
Other Services	5.999	0.056	0.013
Real Estate	1.190	0.028	0.006
Retail Trade	7.527	0.107	0.009
Technical Services	1.043	0.030	0.006
Transportation	2.981	0.037	0.013
Utilities	0.213	0.006	0.002
Wholesale Trade	3.977	0.104	0.030

Table C1: Summary Statistics

This table shows the distribution of Mexican workers across industries and the variation of employment shocks across locations for each industry. Column 1 shows the national employment distribution of Mexicans workers across industries. Column 2 shows variation in the share of Mexican-born workers across commuting zones working in each specific industry in 2006, while column 3 shows the geographic variation in shocks in each industry.

C.2 Subsets of controls

Table C2 shows results paralleling those in Table 4 but including controls for changes in the share of the population with different demographic and educational characteristics. Because the outcomes in Table 4 are measured using full-count tabulations at the municipio level rather than individual-level micro data, the appropriate way to control for potential changes in labor force composition is to include controls for changes in the shares of the local population with each characteristic. As shown in C2, the results in Table 4 are robust to including these compositional controls.

Tables C3 - C7 show results paralleling those in Tables 1 - 6, with different subsets of controls, as in Table 1. We discuss these results in detail in Appendix C.7 in the context of how the inclusion of pre-Recession controls influence the main findings.

	\triangle EPOP	$\Delta \ln(\text{Hours})$	$\triangle \ln(\text{Earnings})$	$\Delta \ln(\text{EarnPerHour})$
	2004-09	2004-09	2004-09	2004-09
	(1)	(2)	(3)	(4)
Panel A. All				
US Employment Shock	-0.172^{*}	-2.488*	-2.204	0.549
	[0.093]	[1.390]	[2.647]	[1.884]
	(0.119)	(1.288)	(2.638)	(2.362)
\triangle EPOP	-0.084*			
1999-04	(0.044)			
$\Delta \ln(\text{Hours})$		-0.113***	0.311***	0.437***
1999-04		(0.38)	(0.113)	(0.098)
$\triangle \ln(\text{Earnings})$		0.060***	-0.200***	-0.259***
1999-04		(0.020)	(0.059)	(0.048)
Mean raw outcome	0.013	0.126	0.284	0.213
among less affected				
Implied shock impact	0.003	0.047	0.041	-0.010
	Ver	Vee	V	Ver
State FE	Yes	Yes	Yes	Yes
Observations	865	846	846	846

Table C2: Employment-to-population Ratio and Earnings per Hour

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio, earnings, hours worked, and earnings per hours in each municipio, using employment, earnings and hours from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. We trim the bottom and top 1 percent of the earnings distribution. Columns (3) and (4) additionally include controls for changes in the share of the working age population who is female, by education level, age category, and employed in each industry from the 2005 and 2010 Mexican Census and the 20014 and 2009 Mexican Economic Census. All specifications in columns (1) to (4) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, changes in homicide rates across municipios, and Mexican state fixed effects. "Mean raw outcome among less affected" is the average of the dependent variable without dividing by exposure for municipios in the quartile with smallest magnitude US employment shocks. "Implied shock impact" provides the predicted difference in the outcome (without dividing by exposure) for municipios with the 90-10 percentile difference in shock size (0.075) and average exposure (0.25). Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Due to the shift-share structure of the US Employment Shock, we also present Borusyak et al. (2022) standard errors for this variable in square brackets. *** p<0.01, ** p<0.05, * p<0.1 based on standard errors in brackets when present.

		\triangle Sex Ratio (F/M) 2005-10	10 10	\Box L	\triangle Less than Primary 2005-10	mary	\triangle P	\triangle Primary 2005-10	5-10	\triangle Se	\triangle Secondary 2005-10	05-10	ΔU	\triangle University 2005-10	05-10	Rei	Household Remittances 2010	010
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)
US Employment Shock	0.689^{***} (0.213)	0.550^{**} (0.192)	0.479^{**} (0.200)	0.204^{*} (0.119)	0.061 (0.102)	-0.019 (0.115)	0.303^{*} (0.159)	0.061 (0.159)	0.049 (0.168)	-0.155^{*} (0.084)	$0.002 \\ (0.071)$	-0.028 (0.076)	-0.193^{***} (0.053)	-0.002 (0.053)	0.030 (0.062)	0.452^{***} (0.129)	0.399^{***} (0.113)	0.470^{***} (0.125)
Pre-shock Outcome		-0.228^{***} (0.032)	-0.224^{***} (0.032)	0.550***	0.541^{***} (0.023)	(0.025)	0.679***	0.713^{***} (0.070)	(0.067)	0.840^{***}	0.838^{***} (0.037)	(0.037)	0.707***	0.705^{***} (0.035)	(0.035)	0.456***	0.447^{***} (0.045)	(0.045)
New 287g Policy			-0.035 (0.048)			-0.082^{***} (0.029)			0.083^{**} (0.037)			-0.017 (0.018)			-0.001 (0.014)			0.003 (0.027)
Employment Policy			$\begin{array}{c} 0.010 \\ (0.037) \end{array}$			0.009 (0.021)			-0.006 (0.029)			-0.009 (0.012)			0.006 (0.011)			0.022 (0.018)
Trade Shock			-0.074^{***} (0.027)			-0.023^{*} (0.013)			0.290^{***} (0.039)			-0.180^{***} (0.030)			-0.017 (0.022)			0.044^{**} (0.020)
Non-tradable share of Employment	<i>c</i>		$\begin{array}{c} 0.015 \\ (0.022) \end{array}$			0.002 (0.013)			0.080^{***} (0.017)			-0.008 (0.010)			-0.030^{***} (0.007)			0.023^{*} (0.013)
Homicide Rate 2005-10			10.482^{***} (3.664)			3.294^{**} (1.641)			5.139^{***} (1.977)			0.401 (0.866)			-1.032 (0.966)			-4.623^{*} (2.478)
State FE Observations R-squared	Yes 866 0.199	Yes 866 0.278	Yes 866 0.294	Yes 866 0.214	Yes 866 0.732	Yes 866 0.733	Yes 866 0.314	Yes 866 0.554	Yes 866 0.590	Yes 866 0.352	Yes 866 0.706	Yes 866 0.713	Yes 866 0.267	Yes 866 0.640	Yes 866 0.648	Yes 866 0.183	Yes 866 0.332	Yes 866 0.343

Table C3: Sex Ratio, Attainment, and Households Receiving Remittances

Census or 2005 Inter-Censal Count. We calculate the share of households receiving remittances as the number of households reporting income from relatives living the sex ratio in a municipio facing the 90th percentile shock fell by 0.009 more than in a municipio at the 10th percentile. The results in columns (4)-(15) show no statistically significant relationship between the shock and the share of population with any particular level of education. Column (18) shows a positive relationship between the share of The point estimate of 0.5 in column (18) implies that, a strongly affected municipio experienced nearly a 1 percentage point larger decline in the share of households receiving remittances compared to a less affected municipio. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are This table examines the effect of changes in US labor demand on the 2005-10 change in the female-to-male sex ratio for the working age population (15-64), the 2005-10 change in the share of the working age population in each education level for each Mexican source municipio, and the share of households receiving remittances in 2010. Note that outcome and pre-shock outcome variables are divided by exposure as in equation (7). We measure the sex ratio and educational attainment using the 2000 or 2010 Mexican abroad divided by the municipio's total number of households in the Census year, using the 2000 or 2010 Mexican Census. All specifications in columns (1) to (18) use a GLS re-weighting procedure to address potential heteroskedasticity. The positive coefficient estimate in column (3) for the US employment shock implies that municipios facing larger US employment declines had larger declines in the sex ratio, consistent with return migrants being disproportionately male. The coefficient of 0.483 in column (3), implies that households receiving remittances and the US employment shock, meaning that remittances fall significantly in Mexican source regions facing larger declines in US labor demand. shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

		$\triangle \text{ EPOP } 2004-09$	4-09	\square	Δ ln(Hours) 2004-09	04-09	$\bigtriangleup \ln($	$\Delta \ln(\text{Earnings}) 2004-09$	2004-09	$\Delta \ln(E)$	$\Delta \ln(\text{EarnPerHour}) 2004-09$:) 2004-09
	(1)	(2)	(3)	(4)	(2)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock	-0.195 (0.169)	-0.292* (0.159)	-0.366^{**} (1.243)	-2.254^{*} (1.234)	-2.461^{**} (1.459)	-2.511*(2.515)	-2.463 (2.425)	-2.387 (2.790)	-2.763 (2.042)	-0.368 (1.873)	-0.120 (2.293)	-0.266
△ EPOP 1999-04		-0.517^{***} (0.117)	-0.511^{***} (0.116)									
$\triangle \ln(\mathrm{Hours})$ 1999-04					-0.126^{***} (0.044)	-0.159^{***} (0.046)		0.307^{**} (0.121)	0.262^{**} (0.122)		0.453^{***} (0.100)	0.435^{***} (0.100)
∆ ln(Earnings) 1999-04					0.059^{**} (0.022)	0.071^{***} (0.023)		-0.187^{***} (0.063)	-0.175^{***} (0.062)		-0.249^{***} (0.050)	-0.247^{**} (0.049)
New 287g Policy			-0.004 (0.052)			0.562 (0.465)			2.081^{**} (0.823)			$1.794^{***} \\ (0.668)$
Employment Policy			-0.019 (0.243)			-0.202 (0.479)			-1.024^{**} (0.394)			-0.876**
Trade Shock			-0.361^{***} (0.085)			-0.575 (0.367)			0.145 (0.781)			0.769 (0.478)
Non-tradable share of Employment			0.073^{**} (0.032)			0.610^{**} (0.246)			0.981^{*} (0.505)			0.429 (0.341)
Homicide Rate 05-10			1.390 (2.671)			-24.375 (19.299)			-76.937^{*} (40.389)			-44.905 (29.723)
State FE Observations R-squared	$\substack{\text{Yes}\\865}\\0.077$	$\substack{\text{Yes}\\865}{0.302}$	$\substack{\text{Yes}\\865}{0.325}$	Yes 848 0.129	Yes 846 0.143	$\substack{\text{Yes}\\846\\0.160}$	Yes 848 0.112	$\substack{\text{Yes}\\846\\0.139}$	Yes 846 0.159	Yes 848 0.070	Yes 846 0.157	$\begin{array}{c} \mathrm{Yes} \\ 846 \\ 0.175 \end{array}$

Table C4: Employment-to-population Batio and Earnings per Hour (All Workers)

earnings distribution. All specifications in columns (1) to (12) use a GLS re-weighting procedure to address potential heteroskedasticity. The negative and significant coefficient estimate for the US employment shock in column (3) implies that regions facing larger US employment declines exhibited larger increases in the employment to population This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio, earnings, hours worked, and earnings per hour in each municipio, using employment, earnings, and hours from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the ratio. Columns (6) and (9) show that both total hours worked and total earnings increased in the most affected municipios. However, column (12) shows no effect on earnings per hour, implying that local labor markets adjusted to the increase in supply of hours worked without reducing wages. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1.

	Δ	EPOP Wo	omen		\triangle EPOP M	len
		2004-09			2004-09	
	(1)	(2)	(3)	(4)	(5)	(6)
US Employment Shock	-0.215	-0.411**	-0.563***	-0.160	-0.176	-0.112
r y i i i	(0.159)	(0.159)	(0.177)	(0.220)	(0.192)	(0.236)
\triangle EPOP		-0.584***	-0.577***		-0.410***	-0.408***
1999-04		(0.159)	(0.159)		(0.089)	(0.087)
New 287g Policy			-0.059			0.061
			(0.045)			(0.074)
Employment Policy			-0.044			0.024
Employment Foncy			(0.032)			(0.043)
Trade Shock			-0.008			-0.737***
			(0.107)			(0.106)
Non-tradable share			0.040			0.109***
of Employment			(0.032)			(0.041)
Homicide Rate 05-10			-0.325			3.581
			(2.506)			(3.628)
State FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	865	865	866	865	865
R-squared	0.098	0.315	0.319	0.058	0.251	0.314

Table C5:	Employment-to-	population	Ratio	bv	Gender

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio for women and men in each municipio, using employment from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (6) use a GLS re-weighting procedure to address potential heteroskedasticity. The coefficient of -0.663 in column (3) implies that a strongly affected municipio with average exposure to the US experienced a 1.2 percentage point larger increase in employment to population ratio among women compared to a similar municipio that was less affected. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

	7	\triangle Computer 2005-10	r	5 0	△ Washing Machine 2005-10	achine	1	△ Ketrigerator 2005-10	tor		Δ TV 2005-10	
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock	-0.360^{**} (0.145)	-0.022 (0.085)	-0.122 (0.101)	$\begin{array}{c} 0.031 \\ (0.215) \end{array}$	$\begin{array}{c} 0.171 \\ (0.198) \end{array}$	0.057 (0.226)	-0.197 (0.232)	0.233 (0.198)	0.211 (0.225)	0.121 (0.178)	0.453^{**} (0.165)	0.474^{**} (0.184)
∆ in Outcome 2000-05		0.876^{***} (0.022)	0.873^{***} (0.022)		0.294^{***} (0.020)	0.279^{***} (0.021)		0.341^{***} (0.020)	0.337^{***} (0.022)		0.394^{***} (0.026)	0.400^{**} (0.027)
New 287g Policy			-0.030 (0.056)			0.064 (0.057)			0.052 (0.048)			0.012
Employment Policy			-0.036 (0.023)			-0.079^{**}			-0.028 (0.041)			$\begin{array}{c} 0.012 \\ (0.033) \end{array}$
Trade Shock			-0.057 (0.040)			0.025 (0.056)			0.057 (0.041)			0.044 (0.031)
Non-tradable share of Employment			-0.019 (0.029)			-0.045 (0.037)			-0.011 (0.030)			0.012
Homicide Rate 05-10			-0.920 (1.516)			0.512 (2.906)			1.418 (2.778)			5.125^{*} (2.820)
State FE Observations R-squared	Yes 866 0.331	Yes 866 0.847	Yes 866 0.848	Yes 866 0.446	Yes 866 0.576	$\begin{array}{c} \mathrm{Yes}\\ 866\\ 0.580 \end{array}$	Yes 866 0.344	Yes 866 0.545	Yes 866 0.546	Yes 866 0.230	Yes 866 0.496	Yes 866 0.500

Table C6: Appliance Ownership

and televisions). We calculate the change in US labor demand on the 2005-10 change in ownership of household durables (personal computers, washing machines, refrigerators, and televisions). We calculate the change in the share households owning the relevant household durable using the 2000 or 2010 Mexican Census or 2005 Conteo. All specifications in columns (1) to (12) use a GLS re-weighting procedure to address potential heteroskedasticity. The coefficient in column (12) is positive and statistically significant, implying that households in more negatively affected municipios slowed down their purchases of televisions. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1This ta

		ttendance es 6-12) 20			Attendance es 13-15) 20			Attendance ges 16-18) 20	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
US Employment Shock	0.143^{**} (0.066)	0.137^{**} (0.067)	0.200^{***} (0.070)	-0.016 (0.205)	0.107 (0.204)	$\begin{array}{c} 0.371 \\ (0.246) \end{array}$	-0.280 (0.205)	-0.078 (0.206)	0.011 (0.241)
\triangle in Outcome 2000-05		-0.013 (0.024)	-0.016 (0.024)		0.126^{***} (0.027)	$\begin{array}{c} 0.114^{***} \\ (0.028) \end{array}$		$\begin{array}{c} 0.193^{***} \\ (0.026) \end{array}$	$\begin{array}{c} 0.176^{***} \\ (0.027) \end{array}$
New 287g Policy			$0.025 \\ (0.017)$			$0.086 \\ (0.058)$			$0.035 \\ (0.058)$
Employment Policy			0.013 (0.012)			0.053 (0.040)			0.002 (0.038)
Trade Shock			0.040^{***} (0.013)			$\begin{array}{c} 0.115^{***} \\ (0.027) \end{array}$			-0.010 (0.037)
Non-tradable share of Employment			-0.008 (0.008)			-0.055^{**} (0.026)			-0.086*** (0.030)
Homicide Rate 05-10			-0.957 (1.176)			-7.930^{**} (4.008)			-3.250 (4.501)
State FE Observations R-squared	Yes 866 0.085	Yes 866 0.086	Yes 866 0.096	Yes 866 0.269	Yes 866 0.288	Yes 866 0.305	Yes 866 0.269	Yes 866 0.323	Yes 866 0.330

Table C7: School Attendance

This table examines the effect of changes in US labor demand on the 2005-10 change in school attendance. We calculate the change in the share of the population in primary (age 6-12), secondary (13-15), and high-school (16-18) reporting having attended school using the 2000 or 2010 Mexican Census or 2005 Conteo. All specifications in columns (1) to (9) use a GLS re-weighting procedure to address potential heteroskedasticity. The coefficient of 0.25 in column (3) implies that a municipio facing a 6.5 percentage point decline in US labor demand experienced 0.4 percentage point larger declines in school attendance for primary school children. This means that in municipios experiencing larger US employment declines, school attendance for primary school children fell more relative to municipios facing smaller employment shocks. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

C.3 Shock variation conditional on exposure

Figure C1 shows a scatter plot relating the US employment shock to exposure, ξ_s , for the municipios in our sample. Although the two are positively related, with modestly higher average US employment shocks in municipios with higher exposure to the US labor market, the extensive variability in US employment shock within narrow ranges of exposure is clearly visible in the scatter plot. The R-squared for a linear regression relating the two quantities is only 0.019.

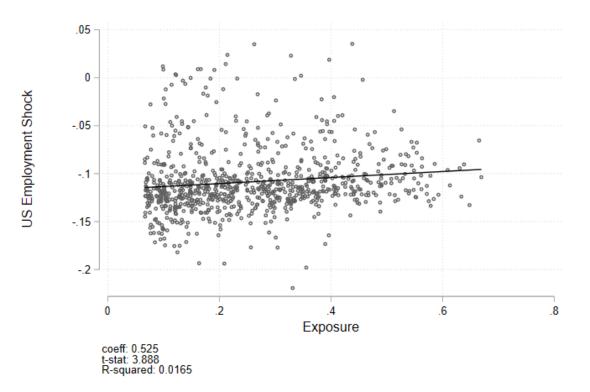


Figure C1: Exposure vs. US Employment Shock Measure

This figure shows the relationship between exposure to the US labor market and the UUS Employment Shock measure across Mexican *municipios*.

C.4 Unweighted Analysis

Tables C8 - C14 show results paralleling those in the main text without the GLS weighting procedure used to address potential heteroskedasticity resulting from dividing the dependent variable by the municipio's estimated exposure to the US labor market. In nearly all cases, we reject the null hypothesis of homoskedastic errors at standard levels using a Breusch-Pagan test, the weighted and unweighted point estimates are very similar, and the weighted standard errors are smaller than the unweighted ones.

	Pop	ulation Gro 2005-10	owth	Re	turn Migrat 2005-10	tion		Emigratic 2005-10	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
US Employment Shock	-3.331^{***} (0.869)	-0.568 (0.812)	-0.473 (0.858)	-0.236^{***} (0.086)	-0.296^{***} (0.074)	-0.213^{***} (0.079)	0.189 (0.127)	$0.196 \\ (0.125)$	0.261^{*} (0.134)
Pre-shock Outcome		$\begin{array}{c} 0.848^{***} \\ (0.177) \end{array}$	$\begin{array}{c} 0.834^{***} \\ (0.175) \end{array}$		1.693^{***} (0.274)	$\begin{array}{c} 1.673^{***} \\ (0.273) \end{array}$		$\begin{array}{c} 0.116^{***} \\ (0.027) \end{array}$	0.108^{***} (0.027)
New 287g Policy			$0.105 \\ (0.193)$			0.077^{***} (0.020)			0.035 (0.028)
Employment Policy			-0.071 (0.114)			-0.012 (0.013)			-0.003 (0.019)
Trade Shock			-0.203 (0.270)			0.020^{**} (0.009)			0.053^{***} (0.016)
Non-tradable share of Employment			-0.284^{**} (0.121)			-0.012 (0.008)			0.007 (0.015)
Homicide Rate 05-10			-25.525^{*} (13.760)			-4.995^{***} (1.177)			-7.303^{***} (2.686)
State FE Observations R-squared	Yes 866 0.143	Yes 866 0.634	Yes 866 0.639	Yes 866 0.302	Yes 866 0.518	Yes 866 0.540	Yes 866 0.246	Yes 866 0.272	Yes 866 0.287
B-P het. p-val	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000

Table C8: Population Growth, Return Migration, and Emigration	(unweighted)
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This table examines the effect of changes in US labor demand on the 2005-10 population growth, return migration to, and emigration from each Mexican source municipio. Note that outcome and pre-shock outcome variables are divided by exposure as in equation (7). We restrict attention to individuals age 15-64. Population growth is defined as the proportional change in population. Return migration is the number of individuals reporting living in the US 5 years prior to the relevant survey, divided by the municipio population in the survey year, while emigration is the number of household members who left for the US during the 5 years prior to the relevant survey, divided by the initial municipio population, measured using the roughly 10% long-form sample from the 2000 or 2010 Census (emigration information is not available in 2005). We use full-count tabulations from the 2000 or 2010 Mexican Census or 2005 Conteo to calculate population growth and return migration. The "Pre-shock Outcome" controls in columns (2), (5), and (8) are 2000-2005 population growth, 2000-2005 return migration, and 1995-2000 emigration, respectively. Columns (3), (6), and (9) additionally control for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios. All specifications control for Mexican state fixed effects, and standard errors clustered at the Mexican commuting zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

(unweighted)
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C9:
Table

	Pop	Population Growth 2005-10	wth	Ret	Return Migration 2005-10	ion		Emigration 2005-10			Net Migration 2005-10	UO	Pop in ar	Population aging in and out 2005-10	ng -10		Residual 2005-10		
	(1)	(2)	(3)	(4)	(2)	(9)	(2)	(8)	(6)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)	
US Employment Shock		-0.039	0.169	-0.236***	-0.250***	-0.174**	0.189	0.211*	0.280^{**}	-0.126	0.511	0.933	-2.694***	-0.178	-0.216	-0.086	0.089	-0.093	
Population Growth 2000-05	(0.809)	(0.737^{***}) (0.259)	(0.728^{***}) (0.257)	(0.080)	(0.007* -0.007* (0.004)	(0.007^{*}) -0.007* (0.004)	(1717)	(0.115) - 0.023^{***} (0.005)	$(0.123) - 0.023^{***}$ (0.005)	(0.403)	(0.385^{**}) (0.191)	(0.004) 0.386^{**} (0.192)	(067.0)	$\begin{pmatrix} 0.142\\ 0.012\\ (0.010) \end{pmatrix}$	$\begin{pmatrix} 0.104 \\ 0.012 \\ (0.010) \end{pmatrix}$	(100.0)	$(0.023) \\ 0.324^{***} \\ (0.082)$	$\binom{0.723}{0.315^{***}}$ $\binom{0.079}{0.079}$	
Return Migration Rate 2000-05		-3.687^{*} (1.933)	-3.670^{*} (1.891)		1.231^{***} (0.284)	1.220^{***} (0.282)		1.073^{***} (0.337)	1.051^{***} (0.326)		-2.041 (1.532)	-2.117 (1.517)		0.122 (0.389)	0.131 (0.395)		-1.926 (1.645)	-1.853 (1.658)	
Emigration Rate 1995-00 Net Migration Rate 2000-05		$\begin{array}{c} 0.724^{***} \\ (0.252) \\ 0.115^{**} \\ (0.050) \end{array}$	$\begin{array}{c} 0.710^{***} \\ (0.246) \\ 0.111^{**} \\ (0.046) \end{array}$		$\begin{array}{c} 0.167^{***} \\ (0.038) \\ 0.011^{**} \\ (0.006) \end{array}$	$\begin{array}{c} 0.161^{***} \\ (0.037) \\ 0.013^{**} \\ (0.006) \end{array}$		$\begin{array}{c} 0.064^{*} \\ (0.033) \\ 0.005 \\ (0.004) \end{array}$	$\begin{array}{c} 0.058^{*} \\ (0.032) \\ 0.006 \\ (0.004) \end{array}$		$\begin{array}{c} 0.478^{***} \\ (0.178) \\ 0.791^{***} \\ (0.176) \end{array}$	$\begin{array}{c} 0.445^{**} \\ (0.177) \\ 0.796^{***} \\ (0.176) \end{array}$		$\begin{array}{c} -0.058\\ (0.044)\\ 0.012\\ (0.009) \end{array}$	$\begin{array}{c} -0.055\\ (0.044)\\ 0.012\\ (0.008) \end{array}$		$\begin{array}{c} 0.201 \\ (0.174) \\ -0.695^{***} \\ (0.202) \end{array}$	$\begin{array}{c} 0.217 \\ (0.177) \\ -0.703^{***} \\ (0.198) \end{array}$	
Population aging in and out 2000-05		0.273^{*} (0.151)	0.268^{*} (0.156)		0.018^{***} (0.005)	0.015^{**} (0.005)		0.048^{***} (0.009)	0.047^{***} (0.010)		-0.151 (0.116)	-0.160 (0.119)		0.893^{***} (0.016)	0.894^{***} (0.016)		-0.439^{***} (0.071)	-0.433^{***} (0.072)	
New 287g Policy			-0.047 (0.197)			0.067^{***} (0.018)			0.012 (0.027)			0.234 (0.165)			-0.028 (0.034)			-0.308^{*} (0.167)	
Employment Policy			0.073 (0.127)			-0.005 (0.012)			0.013 (0.020)			0.030 (0.109)			0.001 (0.027)			0.060 (0.109)	
Trade Shock			-0.202 (0.264)			0.023^{**} (0.009)			0.043^{**} (0.019)			0.145 (0.185)			-0.021 (0.014)			-0.306^{***} (0.111)	
Nontradable			-0.240^{*} (0.127)			-0.011 (0.008)			0.016 (0.013)			-0.026 (0.088)			-0.009 (0.017)			-0.178 (0.110)	
Homicide Rate 2005-10			-20.664 (13.727)			-4.009^{***} (1.047)			-6.493^{***} (1.847)			-26.535^{***} (9.103)			2.664 (2.468)			0.724 (15.790)	
Observations R-squared	866 0.143	866 0.660	866 0.663	866 0.302	866 0.605	866 0.620	866 0.246	866 0.358	866 0.370	866 0.051	866 0.605	866 0.609	866 0.334	866 0.979	866 0.979	866 0.115	866 0.492	866 0.498	866
This table examines the effect of changes in US labor demand on the 2005-10 population growth and its components, including: return migration to, and emigration from each Mexican source municipio, internal net migration, and population aging in and out of the working age population. Note that outcome and pre-shock outcome variables are divided by exposure as in constion (7). We restrict attention to individuals are 15-64. Population is defined as the proportional change in population. Return	the effec municipic ure as in	t of chan , interna equation	ges in Ut l net mig (7). We	5 labor de ration, an restrict a	mand on d populat ttention t	the 2005- Jion aging o individ	-10 popu ; in and uals age	2005-10 population growth and its components, including: return migration to, and emigration from aging in and out of the working age population. Note that outcome and pre-shock outcome variables udividuals age 15-64. Population growth is defined as the proportional change in population. Return	owth and s working opulation	its com age pop growth	oonents, vulation. is define	including: Note that d as the r	: return n t outcome proportion	nigration and pre- nal chang	to, and shock ou e in pop	emigrati utcome v ulation.	on from ⁄ariables Return		

migration is the number of individuals reporting living in the US 5 years prior to the relevant survey, divided by the municipio population in the survey year, while emigration is the number of household members who left for the US during the 5 years prior to the relevant survey, divided by the initial municipio population, measured using the roughly 10% long-form sample from the 2000 or 2010 Census (emigration information is not available in 2005). Internal net migration at the municipio level is calculated as using the roughly 10% long-form sample from the 2000 or 2010 Census (internal migration information is not available in 2005). We use full-count tabulations from the 2000 or 2010 Mexican Census or 2005 Conteo to calculate population growth, return migration, and population aging in and out. The "Pre-shock Outcome" controls in columns (2), (5), (8), (11), (14) are 2000-2005 population growth, 2000-2005 return migration, 1995-2000 emigration, 1995-2000 net migration, and 2000-2005 population aging in and out, respectively. All specifications control for Mexican state fixed effects, and standard errors clustered at the Mexican commuting zone level are shown in parentheses. Standard the difference between each municipio's immigrants and emigrants during the 5 years prior to the to the relevant survey, divided by the initial municipio population, measured errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1 E ea ar

		\triangle Sex Ratio (F/M) 2005-10		$\triangle Le$	△ Less than Pri 2005-10	Primary	$\bigtriangleup \mathrm{P}$	\triangle Primary 2005-10)5-10	\triangle Se	\triangle Secondary 2005-10	05-10	Δ Uı	\triangle University 2005-10	05-10	Rer	Household Remittances 2010	010
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)
US Employment Shock	0.763^{***} (0.235)	0.644^{***} (0.219)	0.589^{**} (0.253)	0.942^{***} (0.293)	0.174 (0.155)	$\begin{array}{c} 0.131 \\ (0.177) \end{array}$	-0.215 (0.361)	0.003 (0.269)	0.159 (0.283)	-0.450^{**} (0.200)	0.117 (0.124)	$\begin{array}{c} 0.010 \\ (0.139) \end{array}$	-0.336^{**} (0.159)	-0.048 (0.093)	-0.018 (0.110)	0.504^{***} (0.137)	0.428^{***} (0.120)	0.510^{***} (0.133)
Pre-shock Outcome		-0.177***(0.035)	-0.175^{***} (0.035)		0.586^{***} (0.025)	0.583^{***} (0.026)	2	0.856^{***} (0.060)	0.865^{***} (0.061)		0.899^{***} (0.046)	0.907^{***} (0.047)		0.780^{***} (0.066)	0.778^{***} (0.067)		0.459^{***} (0.046)	0.450^{***} (0.046)
New 287g Policy			-0.083 (0.072)			-0.080^{**} (0.040)			0.160^{***} (0.061)			-0.093^{**} (0.039)			-0.019 (0.030)			0.008 (0.028)
Employment Policy			0.043 (0.044)			0.019 (0.028)			0.022 (0.040)			-0.000 (0.022)			0.016 (0.019)			0.024 (0.018)
Trade Shock			-0.075**(0.030)			-0.033 (0.031)			0.311^{***} (0.069)			-0.186*** (0.033)			-0.020 (0.029)			0.048^{**} (0.023)
Non-tradable share			0.035 (0.027)			-0.002 (0.017)			0.109^{***} (0.030)			-0.003 (0.019)			-0.046^{***} (0.013)			0.020 (0.014)
Homicide Rate 05-10 of Employment			$10.418^{**} (4.589)$			1.033 (3.103)			$12.770^{***} (4.824)$			-0.139 (1.904)			-1.196 (1.663)			-5.118^{**} (2.505)
State FE Observations R-squared B-P het. p-val	Yes 866 0.197 0.000	Yes 866 0.251 0.000	Yes 866 0.262 0.000	Yes 866 0.315 0.000	Yes 866 0.792 0.000	Yes 866 0.793 0.000	Yes 866 0.298 0.000	Yes 866 0.655 0.000	Yes 866 0.672 0.000	Yes 866 0.356 0.000	Yes 866 0.732 0.000	Yes 866 0.740 0.000	Yes 866 0.241 0.000	Yes 866 0.735 0.000	Yes 866 0.739 0.000	Yes 866 0.187 0.000	Yes 866 0.334 0.000	Yes 866 0.343 0.00
This table examines the effect of changes in US labor demand on the 2005-10 change in the female to male sex ratio for the working age population (15-64), the 2005-10 change in the share of the working age population in each education level for each Mexican source municipio, and the share of households receiving remittances in 2010. Note that outcome and pre-shock outcome variables are divided by exposure as in equation (7). We measure the sex ratio and educational attainment using the 2000 or 2010 Mexican Census or 2005 Inter-Censal Count. We calculate the share of households reporting receiving income from relatives living abroad divided by the muncipio's total number of households reporting receiving income from relatives living US employment shock implies that municipio's facing larger US employment declines and the share of population with any particular level of education. Column (18) shows a positive relationship between the shock and the share of population with any particular level of education. Column (18) shows a positive relationship between the Nexican canded the shock, meaning that remittances fall significantly in Mexican Source sciences for the shore of households receiving remittances are the shock and the share of population with any particular level of education. Column (18) shows a positive relationship between the shore and the US employment shock, meaning that remittances fall significantly in Mexican source regions facing larger US labor demand. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level	the effect vorking aç vorking aç contron -Censal C e municip e municip k implies relations.	of changes ge populat ne variable Nount. We io's total r that munic 6)-(15) shc hip betwee shines in Ul	s in US la ion in ea es are div calculat number of cipios faci ow no sta en the shu S labor d	ubor deme ch educat vided by e e the shau f househol ing larger tistically are of hou emand. A	and on th and on th ion level exposure re of hou lds in the US empl significan seholds r All specifi	ee 2005-1(for each as in equ scholds ry Census y oyment d it relatior ecciving cations ir) change Mexican nation (7, ecciving /ear, usin eclines hi iship betv remittan rclude M	in the fer source 1 b. We mo remittan g the 200 ad larger ween the ces and t	male to n municipic easure th ces as th 00 or 2010 declines shock an the US er ate fixed	nale sex ri- and the sex rati e number 0 Mexicar in the sex id the sha nploymen effects. S	atio for t a share of io and ed of house t Census. t ratio, co re of pop t shock, t andard	he workin f househc uucational sholds rep The pos onsistent ulation w meaning errors clu	ig age pop lds receive attainm porting re itive coeff itive tu with retu vith any p that remi istered at	pulation (ing remi- ent using in ficient est ficient est articular the Mex	(15-64), th trances in the 2000 ncome fro ncome fro imate in the being c level of ec all signific ican Com	ne 2005-1 1 2010. N 1 or 2010 2010 m 1 relativ 2010mn (; 1 1 1 1 1 2 1 1 2 0 1 1 2 0 1 1 1 2 0 1 1 1 1	0 change Note that Mexican ves living 3) for the tionately Column Mexican	

	\Box	EPOP 2004-09			MP (amori)m	ZUU4-U9	\bigtriangleup In(1		2004-09	$\Delta \ln(E_6$	△ In(EarnPerHour)) 2004-09
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock	-0.116 (0.686)	-0.307 (0.537)	-0.759 (0.505)	-7.969^{**} (3.829)	-8.472^{**} (3.755)	-9.226^{**} (3.805)	-9.566^{**} (4.726)	-8.264^{*} (4.524)	-9.875^{**} (4.973)	-1.596 (4.209)	0.208 (3.853)	-0.649 (4.279)
△ EPOP 99-04		-0.376^{**} (0.145)	-0.360^{**} (0.145)									
riangle ln(Hours) 99-04					-0.124^{**} (0.061)	-0.146^{**} (0.062)		0.331^{**} (0.148)	0.290^{**} (0.146)		$\begin{array}{c} 0.454^{***} \\ (0.122) \end{array}$	0.436^{***} (0.122)
riangle ln(Earnings) 99-04					0.067^{**} (0.034)	0.078^{**} (0.035)		-0.177^{**} (0.069)	-0.159^{**} (0.069)		-0.244^{***} (0.054)	-0.237^{***} (0.053)
New 287g Policy			-0.236^{*} (0.135)			0.410 (0.755)			2.962^{**} (1.390)			2.552^{***} (0.967)
Employment Policy			-0.052 (0.077)			-0.453 (0.425)			-2.285^{***} (0.865)			-1.833^{***} (0.650)
Trade Shock			-0.374^{***} (0.101)			-0.743 (0.536)			-0.005 (1.046)			0.739 (0.623)
Non-tradable share of employment			0.315^{***} (0.086)			$\begin{array}{c} 1.227^{***} \\ (0.404) \end{array}$			2.229^{***} (0.794)			1.002^{*} (0.558)
Homicide Rate 05-10			-8.203 (8.258)			-46.380 (46.328)			-129.179*(70.689)			-82.799 (58.155)
State FE Observations R-squared B-P het. p-val	Yes 865 0.102 0.000	Yes 865 0.205 0.000	Yes 865 0.244 0.000	Yes 848 0.119 0.000	Yes 846 0.132 0.000	Yes 846 0.149 0.000	Yes 848 0.114 0.000	Yes 846 0.138 0.000	Yes 846 0.161 0.000	Yes 848 0.086 0.000	Yes 846 0.160 0.003	Yes 846 0.175 0.003

		EPOP Wo	men		\triangle EPOP N	/Ien
		2004-09			2004-09	
	(1)	(2)	(3)	(4)	(5)	(6)
US Employment Shock	-0.327 (0.715)	-0.663 (0.483)	-1.126^{**} (0.484)	0.015 (0.688)	-0.072 (0.613)	-0.409 (0.591)
\triangle EPOP 99-04		-0.545^{***} (0.167)	-0.525^{***} (0.157)		-0.286** (0.114)	-0.273^{**} (0.108)
New 287g Policy			-0.269^{**} (0.136)			-0.139 (0.145)
Employment Policy			-0.086 (0.074)			-0.021 (0.084)
Trade Shock			0.028 (0.107)			-0.733^{***} (0.147)
Non-tradable share of Employment			$\begin{array}{c} 0.237^{***} \\ (0.071) \end{array}$			0.366^{***} (0.100)
Homicide Rate 05-10			-12.321 (9.571)			-4.389 (8.620)
State FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	865	865	866	865	865
R-squared	0.128	0.302	0.325	0.082	0.158	0.204
B-P het. p-val	0.000	0.000	0.000	0.000	0.000	0.000

Table C12: Employment-to-population Ratio by Gender (unweighted)

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio for women and men in each municipio, using employment from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. The negative coefficient in colum (3) implies that a strongly affected municipio experienced larger increase in employment to population ratio among women compared to a similar municipio that was less affected. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

		\triangle Computer 2005-10	er	\square	\triangle Washing Machine 2005-10	tchine	\triangleleft	\triangle Refrigerator 2005-10	or		$\Delta ext{ TV} 2005-10$	
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock	-0.689^{*} (0.380)	-0.062 (0.166)	-0.012 (0.204)	-0.657 (0.435)	-0.083 (0.372)	-0.088 (0.434)	-1.070^{**} (0.487)	-0.087 (0.378)	-0.163 (0.443)	-0.636 (0.398)	0.523^{*} (0.292)	0.378 (0.342)
∆ in Outcome 2000-05		0.893^{***} (0.034)	0.893^{***} (0.033)		0.320^{***} (0.026)	$\begin{array}{c} 0.314^{***} \\ (0.027) \end{array}$		0.355^{***} (0.035)	0.354^{***} (0.035)		0.473^{***} (0.036)	0.479^{**} (0.036)
New 287g Policy			0.043 (0.057)			0.208^{**} (0.097)			0.282^{**} (0.122)			0.102 (0.086)
Employment Policy			-0.001 (0.038)			-0.101 (0.067)			-0.153^{*} (0.082)			-0.074 (0.056)
Trade Shock			-0.049 (0.067)			0.125 (0.104)			$\begin{array}{c} 0.120 \\ (0.097) \end{array}$			$\begin{array}{c} 0.016 \\ (0.035) \end{array}$
Non-tradable share of Employment			-0.051^{**} (0.024)			-0.032 (0.045)			0.099^{*} (0.054)			0.098^{**} (0.050)
Homicide Rate 05-10			0.612 (2.657)			3.245 (4.922)			3.897 (5.242)			18.014^{**} (8.736)
State FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	866	866	866	866	866	866	866	866	866	866	866
R-squared B-P het. p-val	0.338 0.000	0.883 0.000	0.883 0.000	$0.574 \\ 0.000$	0.678	0.680 0.000	0.422 0.000	0.599 0.000	0.605 0.000	$0.246 \\ 0.000$	0.559 0.000	0.568 0.000

Table C13: Appliance Ownership (unweighted)

and televisions). We calculate the change in the share households owning the relevant household durable using the 2000 or 2010 Mexican Census or 2005 Conteo. The coefficient in column (12) is positive and statistically significant, implying that households in more negatively affected municipios slowed down their purchases of televisions. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1 This ta

		ttendance s $6-12$) 20			Attendance ges 13-15) 20			Attendance ges 16-18) 20	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
US Employment Shock	0.046 (0.117)	$0.050 \\ (0.115)$	0.127 (0.122)	-0.166 (0.327)	$\begin{array}{c} 0.135 \\ (0.330) \end{array}$	$\begin{array}{c} 0.350 \\ (0.383) \end{array}$	-0.060 (0.431)	$0.431 \\ (0.419)$	$\begin{array}{c} 0.533 \\ (0.463) \end{array}$
\triangle in Outcome 2000-05		0.005 (0.029)	0.004 (0.030)		$\begin{array}{c} 0.162^{***} \\ (0.034) \end{array}$	$\begin{array}{c} 0.152^{***} \\ (0.033) \end{array}$		$\begin{array}{c} 0.286^{***} \\ (0.034) \end{array}$	$\begin{array}{c} 0.272^{***} \\ (0.034) \end{array}$
New 287g Policy			0.004 (0.025)			0.113 (0.094)			0.144 (0.095)
Employment Policy			0.030 (0.021)			0.020 (0.059)			-0.038 (0.068)
Trade Shock			0.049^{**} (0.023)			$\begin{array}{c} 0.104^{***} \\ (0.039) \end{array}$			-0.038 (0.057)
Non-tradable share of Emplyment			-0.008 (0.013)			-0.090^{***} (0.035)			-0.130*** (0.048)
Homicide Rate 05-10			-0.688 (1.888)			-8.664* (4.834)			-5.735 (6.990)
State FE Observations R-squared B-P het. p-val	Yes 866 0.092 0.111	Yes 866 0.092 0.045	Yes 866 0.099 0.021	Yes 866 0.221 0.000	Yes 866 0.269 0.000	Yes 866 0.279 0.000	Yes 866 0.241 0.000	Yes 866 0.358 0.000	Yes 866 0.367 0.000

Table C14: School Attendance (unweighted)

This table examines the effect of changes in US labor demand on the 2005-10 change in school attendance. We calculate the change in the share of the population in primary (age 6-12), secondary (13-15), and high-school (16-18) reporting having attended school using the 2000 or 2010 Mexican Census or 2005 Conteo. The positive coefficient in column (3) implies that in municipios experiencing larger US employment declines, school attendance for primary school children fell more relative to municipios facing smaller employment shocks. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

C.5 Municipios' baseline characteristics by primary US destination

Table C15 examines the baseline characteristics of municipios based on their migrants' primary US destinations, as reported in the MCAS data. We first assign each municipio to one of either California, Illinois, or Texas based on which US state accounts for the largest share of the municipio's US migrants (these three states are by far the most important destinations for Mexican-born migrants in the US). Table C15 then reports average demographic, educational, and economic characteristics for each group of municipios. The observable municipio characteristics are very similar across the three groups, supporting the "exogenous shares" approach to shift-share causal inference proposed by Goldsmith-Pinkham et al. (2020).

	California	Illinois	Texas
Sex	51.529	52.207	50.923
Ages 15-19	10.900	10.469	10.331
Ages 20-24	8.629	8.240	8.357
Ages 25-29	7.209	6.985	7.219
Ages 30-34	6.363	6.228	6.569
Ages 35-39	5.790	5.669	5.843
Ages 40-44	4.799	4.747	4.926
Ages 45-49	3.905	3.920	3.983
Ages 50-54	3.364	0.478	0.570
Ages 55-59	2.755	2.920	0.967
Ages 60-64	2.486	2.682	2.789
Less than primary completed	52.090	54.471	50.716
Primary completed	39.949	38.728	40.797
Secondary completed	5.716	4.835	5.996
University completed	2.246	1.966	2.492
Employed	44.924	43.242	42.269
Unemployed	0.510	0.454	0.474
Not in labor force	54.566	56.304	57.257
Self-employed	32.210	32.285	27.678
Wage/salary worker	58.530	57.310	63.800
Unpaid worker	9.260	10.405	8.521
Total	560	38	164
Observations	762	762	762

Table C15: Municipios' baseline characteristics

This table shows descriptive evidence on the baseline characteristics of *municipios* with primary connections to California, Illinois, and Texas, using data from the 2000 Mexican Census. *Municipios* are assigned to one of the US states in each column based on their largest connection as represented by the migrant network from the MCAS data. The similar distribution of baseline characteristics for each US state suggests that equal counterfactual trends assumption is plausible.

C.6 Mexican Commuting Zone Analysis for Labor Market Outcomes

For consistency with the rest of the analysis, the labor market outcome results in Table 4 use Mexican municipio as the unit of analysis. Here, we provide a parallel analysis using Mexican commuting zones as the unit of analysis, in order to address the possibility that municipios in the same commuting zone may be part of an integrated labor market in equilibrium. We define Mexican commuting zones following Atkin (2016), and impose the same sample restrictions to commuting zones that we did to municipios in the main text: at least 5,000 residents in 2005, exposure $\xi_s > 0.066$ (the 25th percentile), and at least 100 MCAS cards in 2006. This yields a sample of 741 Mexican commuting zones (and 723 with information in the Economic Census). The results in Table C16 are extremely similar to those in Table 4, showing that the choice of Mexican market aggregation does not substantially affect our findings.

	\triangle EPOP	$\Delta \ln(\text{Hours})$	$\triangle \ln(\text{Earnings})$	$\triangle \ln(\text{EarnPerHour})$
	2004-09	2004-09	$\simeq m(Earmigs)$ 2004-09	$\simeq m(Earreferriour)$ 2004-09
	$\frac{2004-09}{(1)}$	(2)	(3)	(4)
	(1)	(2)	(3)	(4)
<u>Panel A. All</u> US Employment Sheel	-0.375**	-3.554***	2 400	0 117
US Employment Shock			-3.422	0.117
A EDOD	(0.184) - 0.534^{***}	(1.323)	(2.672)	(2.199)
\triangle EPOP 1999-04				
1999-04	(0.104)			
$\wedge \ln(\mathbf{H}_{ours})$		-0.183***	0.198	0.399***
$\Delta \ln(\text{Hours})$				
1999-04		(0.049)	(0.131)	(0.104)
$\triangle \ln(\text{Earnings})$		0.054**	-0.174**	-0.231***
1999-04		(0.027)	(0.072)	(0.054)
Panel B. Women		(0:021)	(0.012)	(0.001)
US Employment Shock	-0.547***			
es Employment Shoek	(0.202)			
	(0.202)			
\triangle EPOP	-0.612***			
Women 1999-04	(0.151)			
Wollien 1999 01	(0.101)			
Panel C. Men				
US Employment Shock	-0.260			
es Employment Shoek	(0.215)			
	(0.210)			
\triangle EPOP	-0.409***			
Men 1999-04	(0.088)			
State FE	Yes	Yes	Yes	Yes
Observations	741	723	723	723

Table C16:	Employment-to-pop	pulation Ratio	and Earnings	per Hour

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio, earnings, hours worked, and earnings per hours in each Mexican commuting zone level, using employment, earnings and hours from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (4) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, changes in homicide rates across municipios, and Mexican state fixed effects. Robust standard errors (equivalent to clustering at the Mexican commuting zone level) are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

C.7 Pre-shock placebo tests

As discussed in Section 5.1.4 in the main text, an analysis of the relationship between the pre-shock values of the outcomes we study and the future shock that municipios eventually experience aids in the interpretation of our analysis. In the same way that Table 2 provides pre-trend analysis for Table 1, Tables C17 – C21 provide pre-trend analysis for the outcomes from Tables C29 – 6.

Table C17 shows the placebo tests corresponding to the pre-shock outcome controls in Table 3. It is also helpful to compare the placebo test results to Table C3, which presents the main analysis with and without the pre-Recession controls to see how their inclusion affects the results. We find meaningful pre-Recession relationships for the change in sex ratio, and the primary, secondary, and university educational attainment shares. Introducing these pre-Recession controls in to Table C3 has a nontrivial effect on the educational attainment coefficients. While the pre-Recession control for the sex ratio does reduce the magnitude of the contemporaneous coefficient somewhat, it does not qualitatively change the conclusion. Importantly, there is no evidence of a pre-Recession relationship for household remittances.

Table C18 shows the placebo tests corresponding to the pre-shock outcome controls in Panel A of Table 4. It is also helpful to compare the placebo test results to Table C4, which presents the main analysis with and without the pre-Recession controls to see how their inclusion affects the results. We find minimal sign of pre-Recession outcome relationships in Table C18 for employment, hours, and earnings outcomes, which is consistent with the stable coefficients across columns in Table C4.

Table C19 shows the placebo tests corresponding to the pre-shock outcome controls in Panel B of Table 4, separately by gender. It is also helpful to compare the placebo test results to Table C5, which presents the main analysis with and without the pre-Recession controls to see how their inclusion affects the results. While there is little sign of a pre-Recession relationship for the employment-to-population ratio for men, there is an apparent relationship for women. Nonetheless, the inclusion of the pre-Recession control in Table C5 does not qualitatively change the relationship between the change in women's employment-to-population ratio and the US employment shock faced by their municipio, although it does increase the precision of the estimates.

Table C20 shows the placebo tests corresponding to the pre-shock outcome controls in columns (1)-(4) of Table 6, examining appliance ownership. It is also helpful to compare the placebo test results to Table C6, which presents the main analysis with and without the pre-Recession controls to see how their inclusion affects the results. Table C20 finds nontrivial pre-Recession relationships between the change in ownership and subsequent US employment shocks for all appliances. This pattern corresponds to the nontrivial changes in the coefficient estimates in Table C6 when introducing the pre-Recession controls. We therefore interpret the results on Television ownership in Table 6 with caution.

Table C21 shows the placebo tests corresponding to the pre-shock outcome controls in columns (5)-(7) of Table 6, examining schooling attendance rates. It is also helpful to compare the placebo test results to Table C7, which presents the main analysis with and without the pre-Recession controls to see how their inclusion affects the results. Table C21 finds meaningful pre-Recession relationships for the change in attendance rate among all age groups, but these have the opposite sign of our main results. Also, Table C7 shows that, particularly for the statistically significant effect among children age 6-12, the inclusion of the pre-Recession outcome growth has minimal effect on the estimates.

Finally, Table C22 provides analysis similar to the cross-sectional specifications in Table 5 in the main text but using data from 2000 rather than from 2010. It shows that, in some specifications, the negative coefficient on the interaction of the US Employment shock (during the Great Recession) and the household exposure indicator existed even prior to the onset of the Recession. These results reinforce the interpretation that the results in Table 5 of the main paper are primarily suggestive evidence of labor supply responses among affected households and should not be treated as definitive.

	\triangle Ses (F/M)	\triangle Sex Ratio (F/M) 2000-05	Δ Less ti 20	Less than Primary 2000-05	$\Delta P_{.}$	∆ Primary 2000-05	$\Delta \operatorname{Sec}_{200}$	Secondary 2000-05	Δ Uni 200(\triangle University 2000-05	Hous Remitta	Household Remittances 2000
		(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock	-0.562^{**} (0.252)	-0.651^{**} (0.271)	0.292^{*} (0.152)	0.359^{**} (0.182)	$\begin{array}{c} 0.197 \\ (0.147) \end{array}$	0.297 (0.181)	-0.239^{***} (0.070)	-0.277^{***} (0.081)	-0.273^{***} (0.068)	-0.337^{***} (0.072)	$\begin{array}{c} 0.145 \\ (0.098) \end{array}$	0.135 (0.113)
New 287g Policy		0.111^{*} (0.061)		-0.081^{*} (0.047)		0.108^{**} (0.039)		0.032 (0.022)		-0.026 (0.019)		-0.066^{*} (0.034)
Employment Policy		-0.092^{**} (0.040)		0.083^{***} (0.030)		-0.021 (0.028)		-0.041^{***} (0.014)		-0.017 (0.013)		0.023 (0.023)
Trade Shock		-0.093^{***} (0.034)		-0.092^{***} (0.033)		-0.067 (0.048)		0.070^{**} (0.024)		0.037 (0.025)		0.022 (0.025)
Non-tradable share of Employment		-0.042 (0.027)		0.095^{***} (0.019)		-0.059^{***} (0.021)		-0.056^{***} (0.011)		-0.005 (0.009)		0.030^{**} (0.013)
Homicide Rate 2005-10		-1.032 (4.181)		4.624 (3.511)		-6.217^{**} (3.016)		-1.716^{*} (0.995)		-0.793 (0.902)		-3.451^{***} (1.326)
State FE Observations R-squared	Yes 866 0.241	$\begin{array}{c} {\rm Yes}\\ 866\\ 0.250 \end{array}$	$\begin{array}{c} \mathrm{Yes}\\ 866\\ 0.210 \end{array}$	Yes 866 0.248	Yes 866 0.218	Yes 866 0.243	$\begin{array}{c} \mathrm{Yes}\\ 866\\ 0.205 \end{array}$	$\begin{array}{c} \mathrm{Yes}\\ 866\\ 0.260 \end{array}$	Yes 866 0.146	Yes 866 0.154	Yes 866 0.296	Yes 866 0.306
This table examines the effect of changes in US labor demand on the pre-shock change in the female to male sex ratio for the working age population (15-64), the change in the share of the working age population in each education level for each Mexican source municipio, and the share of households receiving remittances as a placebo test. Note that the pre-shock outcome variables are divided by exposure as in equation (7). We measure the sex ratio and educational attainment using the 2000 Mexican Census and the 2005 Conteo. We calculate the share of households receiving remeittances as the number of households receiving income from relatives living abroad divided by the municipio's total number of households in the Census year, using the 2000 Mexican Census. All specifications in columns (1) to (12) use a GLS re-weighting procedure to address potential heteroskedasticity and include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. ***	ct of change population variables ar ate the shar r of househo asticity and	es in US labo in each edu e divided by de of househo alds in the C	or demand (cation level ' exposure a lds receivin ensus year, xican state	on the pre-shocl for each Mexic is in equation (g remittances a using the 2000 fixed effects. St	k change i an source 7). We mu is the num Mexican C	in the female municipio, a easure the se aber of house Census. All s rrors clustere	to male sex nd the share x ratio and holds reporti pecifications d at the Mer	ratio for the of household educational a ing receiving in columns (xican Comm	working age ls receiving ru thainment us income from 1) to (12) us thing Zone le	e population - emittances as sing the 2000 t relatives livi ie a GLS re-w svel are shown	(15-64), the s a placebo Mexican (mg abroad reighting pr	e change in test. Note Census and divided by ocedure to heses. ***

Table C17: Placebo test: Sex Ratio, Attainment, and Households Receiving Remittances

	$\triangle EPO$	P 1999-04		ours) 1999-04	$\Delta \ln(\text{Ear})$	rnings) 1999-04	$\Delta \ln(\text{Ear})$	nPerHour) 1999-04
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
US Employment Shock	-0.238 (0.177)	-0.338^{*} (0.183)	-0.719 (1.653)	-1.157 (1.896)	-0.312 (3.365)	-0.385 (3.875)	-0.436 (2.729)	0.072 (2.889)
	(0.111)	(0.100)	(11000)	(1.000)	(0.000)	(0.0.0)	()	(1.000)
New 287g Policy		-0.013		0.348		1.356		0.981
		(0.050)		(0.399)		(0.884)		(0.743)
Employment Policy		-0.043		-0.252		-0.899		-0.557
		(0.035)		(0.267)		(0.694)		(0.557)
Trade Shock		0.102		0.608		1.347*		0.830
		(0.126)		(0.438)		(0.742)		(0.605)
Non-tradable share		-0.043		0.626**		-1.261**		-2.089***
of Employment		(0.046)		(0.286)		(0.624)		(0.498)
Homicide Rate		2.172		-54.690***		-101.643*		-57.548
2005-10		(2.950)		(21.147)		(56.253)		(45.102)
State FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	865	865	848	848	846	846	846	846
R-squared	0.060	0.065	0.116	0.133	0.088	0.106	0.070	0.111

Table C18: Placebo test: Employment-to-population Ratio and Earnings per Hour (All Workers)

This table examines the effects of declines in US labor demand on the pre-shock change in the employment-to-population ratio, earnings, hours worked, and earnings per hour in each municipio as a placebo test. We use employment, earnings, and hours from the 1999 and 2004 Mexican Economic Census and population from the 2000 and the 2005 Conteo. Note that the pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (8) use a GLS re-weighting procedure to address potential heteroskedasticity and include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

	\triangle EPO	P Women	\triangle EPC	P Men
	199	99-04	199	9-04
	(1)	(2)	(3)	(4)
US Employment Shock	-0.375**	-0.413***	-0.004	-0.165
	(0.157)	(0.159)	(0.236)	(0.258)
New 287g Policy		-0.027		0.011
		(0.044)		(0.067)
Employment Policy		-0.012		-0.071
		(0.029)		(0.047)
		0.010		0.110
Trade Shock		0.016		0.118
		(0.173)		(0.109)
Non-tradable share		-0.068*		-0.014
of Employment		(0.039)		(0.058)
or Employment		(0.009)		(0.058)
Homicide Rate 05-10		2.465		2.858
		(2.340)		(4.319)
		```		、
State FE	Yes	Yes	Yes	Yes
Observations	865	865	865	865
R-squared	0.070	0.079	0.053	0.057

Table C19: Placebo test: Employment-to-population Ratio by Gender

This table examines the effects of declines in US labor demand on the pre-shock change in the employment-to-population ratio for women and men in each municipio as a placebo test. We use employment from the 1999 and 2004 Mexican Economic Census and population from the 2000 and the 2005 Conteo. Note that the pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (4) use a GLS re-weighting procedure to address potential heteroskedasticity and include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

		mputer 0-05		ing Machine 00-05		rigerator 10-05		TV 0-05
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
US Employment Shock	-0.454***	-0.508***	-0.516*	-0.721**	-1.263***	-1.221***	-0.952***	-1.235***
1 0	(0.157)	(0.168)	(0.270)	(0.329)	(0.312)	(0.388)	(0.241)	(0.309)
New 287g Policy		-0.049		0.230***		0.336***		0.138*
		(0.052)		(0.088)		(0.097)		(0.071)
Employment Policy		-0.005		-0.222***		-0.233***		-0.213***
		(0.030)		(0.067)		(0.079)		(0.049)
Trade Shock		0.048		0.059		0.131**		$0.085^{*}$
		(0.086)		(0.078)		(0.065)		(0.048)
Non-tradable share		-0.022		-0.174***		-0.229***		-0.133***
of Employment		(0.021)		(0.039)		(0.043)		(0.039)
Homicide Rate		0.119		-8.613*		-24.715***		-8.131*
2005-10		(2.380)		(4.583)		(9.064)		(4.161)
State FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	866	866	866	866	866	866	866
R-squared	0.268	0.272	0.206	0.255	0.305	0.362	0.397	0.425

#### Table C20: Placebo test: Appliance Ownership

This table examines the effect of changes in US labor demand on the pre-shock change in ownership of household durables (personal computers, washing machines, refrigerators, and televisions) as a placebo test. We calculate the change in the share households owning the relevant household durable using the 2000 Mexican Census and the 2005 Conteo. All specifications in columns (1) to (8) use a GLS re-weighting procedure to address potential heteroskedasticity and include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

	$\triangle$ Attend	ance Rate	$\triangle$ Attend	ance Rate	$\triangle$ Attend	ance Rate
	(ages 6-12	2) 2000-05	(ages $13-1$	5) 2000-05	(ages 16-1	8) 2000-05
	(1)	(2)	(3)	(4)	(5)	(6)
US Employment Shock	-0.397***	-0.474***	-0.943***	-0.979***	-0.867***	-0.838***
	(0.097)	(0.110)	(0.262)	(0.268)	(0.256)	(0.291)
New 287g Policy		0.010		-0.098		-0.048
		(0.027)		(0.071)		(0.074)
Employment Policy		-0.035*		-0.007		-0.022
1 0 0		(0.018)		(0.045)		(0.052)
Trade Shock		0.053***		0.146***		0.187**
		(0.016)		(0.053)		(0.082)
Non-tradable share		-0.029**		-0.143***		-0.194***
of Employment		(0.014)		(0.035)		(0.038)
Homicide Rate		0.927		-6.257		-8.459
2005-10		(1.962)		(7.104)		(5.313)
				37	37	3.7
State FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	866	866	866	866	866	866
R-squared	0.236	0.246	0.207	0.235	0.283	0.319

Table C21: Placebo test: School Attendance

This table examines the effect of changes in US labor demand on the pre-shock change in school attendance as a placebo test. We calculate the change in the share of the population in primary (age 6-12), secondary (13-15), and high-school (16-18) reporting having attended school using the 2000 and the 2005 Conteo. All specifications in columns (1) to (6) use a GLS re-weighting procedure to address potential heteroskedasticity and include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

$\begin{array}{ c c c c c c c c c c c c c c c c c c c$		State FE	State FE	Municipio FE
US Employment Shock*1(exposed_h) $-0.103$ (0.163) $-0.000$ (0.096) $-0.121^*$ (0.063) $1(exposed_h)$ $-0.036^{**}$ (0.016) $-0.020^{**}$ (0.009) $-0.010^{**}$ (0.005)Observations $27,969,374$ $27,969,374$ $27,969,374$ $27,969,37427,969,374Panel B. WomenUS Employment Shock*1(exposed_h)-0.201(0.214)-0.088(0.123)-0.293^{***}(0.059)1(exposed_h)-0.044^*(0.024)-0.024^*(0.014)-0.011^{**}(0.005)Observations14,715,94514,715,94514,715,945Panel C. MenUS Employment Shock*1(exposed_h)0.003(0.168)0.098(0.125)0.087(0.072)1(exposed_h)-0.028^{**}(0.014)-0.006(0.007)-0.006(0.007)-0.006(0.007)$		(1)	(2)	(3)
$1 (exposed_h)$ $(0.163)$ $(0.096)$ $(0.063)$ $1 (exposed_h)$ $-0.036^{**}$ $(0.016)$ $-0.020^{**}$ $(0.009)$ $-0.010^{**}$ $(0.005)$ Observations $27,969,374$ $27,969,374$ $27,969,374$ $27,969,374$ $27,969,374$ $Panel B. Women$ US Employment Shock*1(exposed_h) $-0.201$ $(0.214)$ $-0.088$ $(0.123)$ $-0.293^{***}$ $(0.059)$ $1 (exposed_h)$ $-0.044^*$ $(0.024)$ $-0.024^*$ $(0.014)$ $-0.011^{**}$ $(0.005)$ Observations $14,715,945$ $14,715,945$ $14,715,945$ $Panel C. Men$ US Employment Shock*1(exposed_h) $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ $1 (exposed_h)$ $-0.028^{**}$ $(0.014)$ $-0.006$ $(0.007)$ $-0.028^{**}$ $(0.008)$ $-0.006$ $(0.007)$				
$\begin{split} \mathbb{I}(\text{exposed}_h) & \begin{array}{c} -0.036^{**} & -0.020^{**} & -0.010^{**} \\ (0.016) & (0.009) & (0.005) \\ \end{array} \\ \hline \\ \text{Observations} & 27,969,374 & 27,969,374 & 27,969,374 \\ \hline \\ \underline{Panel \ B. \ Women} \\ \text{US Employment Shock*1(exposed}_h) & \begin{array}{c} -0.201 & -0.088 & -0.293^{***} \\ (0.214) & (0.123) & (0.059) \\ \end{array} \\ \hline \\ \mathbb{I}(\text{exposed}_h) & \begin{array}{c} -0.044^{*} & -0.024^{*} & -0.011^{**} \\ (0.024) & (0.014) & (0.005) \\ \end{array} \\ \hline \\ \text{Observations} & 14,715,945 & 14,715,945 & 14,715,945 \\ \hline \\ \underline{Panel \ C. \ Men} \\ \text{US Employment Shock*1(exposed}_h) & \begin{array}{c} 0.003 & 0.098 \\ (0.168) & (0.125) & (0.072) \\ \end{array} \\ \hline \\ \mathbb{I}(\text{exposed}_h) & \begin{array}{c} -0.028^{**} & -0.014^{*} & -0.006 \\ (0.014) & (0.008) & (0.007) \\ \end{array} \end{split}$	US Employment $Shock^* \mathbb{1}(exposed_h)$			
$(C I - u)$ $(0.016)$ $(0.009)$ $(0.005)$ Observations $27,969,374$ $27,969,374$ $27,969,374$ $\frac{Panel B. Women}{US Employment Shock*1(exposed_h)}$ $-0.201$ $(0.214)$ $-0.088$ $(0.123)$ $-0.293^{***}$ $(0.059)$ $1(exposed_h)$ $-0.044^*$ $(0.024)$ $-0.024^*$ $(0.014)$ $-0.011^{**}$ $(0.005)$ Observations $14,715,945$ $14,715,945$ $14,715,945$ $\frac{Panel C. Men}{US Employment Shock*1(exposed_h)}$ $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ $1(exposed_h)$ $-0.028^{**}$ $(0.014)$ $-0.006$ $(0.007)$ $-0.006$ $(0.007)$		(0.163)	(0.096)	(0.063)
$(C I - u)$ $(0.016)$ $(0.009)$ $(0.005)$ Observations $27,969,374$ $27,969,374$ $27,969,374$ $\frac{Panel B. Women}{US Employment Shock*1(exposed_h)}$ $-0.201$ $(0.214)$ $-0.088$ $(0.123)$ $-0.293^{***}$ $(0.059)$ $1(exposed_h)$ $-0.044^*$ $(0.024)$ $-0.024^*$ $(0.014)$ $-0.011^{**}$ $(0.005)$ Observations $14,715,945$ $14,715,945$ $14,715,945$ $\frac{Panel C. Men}{US Employment Shock*1(exposed_h)}$ $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ $1(exposed_h)$ $-0.028^{**}$ $(0.014)$ $-0.006$ $(0.007)$ $-0.006$ $(0.007)$	-1 ( 1 )	0.000**	0.000**	0.010**
Observations $27,969,374$ $27,969,374$ $27,969,374$ Panel B. Women US Employment Shock*1(exposed_h) $-0.201$ (0.214) $-0.088$ (0.123) $-0.293^{***}$ (0.059)1(exposed_h) $-0.044^*$ (0.024) $-0.024^*$ (0.014) $-0.011^{**}$ (0.005)Observations14,715,94514,715,94514,715,945Panel C. Men US Employment Shock*1(exposed_h) $0.003$ (0.168) $0.098$ (0.125) $0.087$ (0.072)1(exposed_h) $-0.028^{**}$ (0.014) $-0.006$ (0.007)	$\mathbb{I}(\operatorname{exposed}_h)$			
$\frac{Panel B. Women}{\text{US Employment Shock*1(exposed}_h)} \begin{array}{c} -0.201 \\ (0.214) \\ (0.123) \\ (0.123) \\ (0.059) \end{array} \\ 1(\text{exposed}_h) \\ \frac{-0.044^* \\ (0.024) \\ (0.014) \\ (0.014) \\ (0.005) \end{array} \\ 0 \\ \frac{Panel C. Men}{\text{US Employment Shock*1(exposed}_h)} \\ 14,715,945 \\ 14,715,945 \\ 14,715,945 \\ 14,715,945 \\ 14,715,945 \\ (0.125) \\ (0.072) \\ 1(\text{exposed}_h) \\ \frac{-0.028^{**} \\ -0.014^* \\ (0.008) \\ (0.007) \\ \end{array} $		(0.016)	(0.009)	(0.005)
$\frac{Panel B. Women}{\text{US Employment Shock*1(exposed}_h)} \begin{array}{c} -0.201 \\ (0.214) \\ (0.123) \\ (0.123) \\ (0.059) \end{array} \\ 1(\text{exposed}_h) \\ \frac{-0.044^* \\ (0.024) \\ (0.014) \\ (0.014) \\ (0.005) \end{array} \\ 0 \\ \frac{Panel C. Men}{\text{US Employment Shock*1(exposed}_h)} \\ 14,715,945 \\ 14,715,945 \\ 14,715,945 \\ 14,715,945 \\ 14,715,945 \\ (0.125) \\ (0.072) \\ 1(\text{exposed}_h) \\ \frac{-0.028^{**} \\ -0.014^* \\ (0.008) \\ (0.007) \\ \end{array} $	Observations	27 969 374	27 969 374	27 969 374
US Employment Shock*1(exposed_h) $-0.201$ (0.214) $-0.088$ (0.123) $-0.293^{***}$ (0.059)1(exposed_h) $-0.044^*$ (0.024) $-0.024^*$ (0.014) $-0.011^{**}$ (0.005)Observations14,715,94514,715,94514,715,945 $\underline{Panel \ C.\ Men}$ US Employment Shock*1(exposed_h) $0.003$ (0.168) $0.098$ (0.125) $0.087$ (0.072)1(exposed_h) $-0.028^{**}$ (0.014) $-0.006$ (0.007) $0.007$ $0.007$	00501 valions	21,909,914	21,303,314	21,303,314
US Employment Shock*1(exposed_h) $-0.201$ (0.214) $-0.088$ (0.123) $-0.293^{***}$ (0.059)1(exposed_h) $-0.044^*$ (0.024) $-0.024^*$ (0.014) $-0.011^{**}$ (0.005)Observations14,715,94514,715,94514,715,945 $\underline{Panel \ C.\ Men}$ US Employment Shock*1(exposed_h) $0.003$ (0.168) $0.098$ (0.125) $0.087$ (0.072)1(exposed_h) $-0.028^{**}$ (0.014) $-0.006$ (0.007) $0.007$ $0.007$	Panel B. Women			
$\begin{array}{c} (0.214) & (0.123) & (0.059) \\ 1(exposed_h) & \begin{array}{c} -0.044^* & -0.024^* & -0.011^{**} \\ (0.024) & (0.014) & (0.005) \end{array} \\ \\ \hline Observations & 14,715,945 & 14,715,945 & 14,715,945 \\ \hline Panel \ C. \ Men \\ US \ Employment \ Shock*1(exposed_h) & \begin{array}{c} 0.003 & 0.098 & 0.087 \\ (0.168) & (0.125) & (0.072) \end{array} \\ \\ 1(exposed_h) & \begin{array}{c} -0.028^{**} & -0.014^* & -0.006 \\ (0.014) & (0.008) & (0.007) \end{array} \end{array}$		-0.201	-0.088	-0.293***
$\begin{split} \mathbb{I}(\text{exposed}_h) & \begin{array}{c} -0.044^* & -0.024^* & -0.011^{**} \\ (0.024) & (0.014) & (0.005) \\ \hline \\ \text{Observations} & 14,715,945 & 14,715,945 & 14,715,945 \\ \hline \\ \frac{Panel\ C.\ Men}{\text{US\ Employment\ Shock}^*\mathbb{I}(\text{exposed}_h)} & \begin{array}{c} 0.003 & 0.098 & 0.087 \\ (0.168) & (0.125) & (0.072) \\ \hline \\ \mathbb{I}(\text{exposed}_h) & \begin{array}{c} -0.028^{**} & -0.014^* & -0.006 \\ (0.014) & (0.008) & (0.007) \\ \end{array} \end{split}$	$\mathbf{r} = \mathbf{r} + $			
$(0.024)$ $(0.014)$ $(0.005)$ Observations14,715,94514,715,94514,715,945 $\underline{Panel C. Men}$ US Employment Shock*1(exposed_h) $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ $1(exposed_h)$ $-0.028^{**}$ $(0.014)$ $-0.004^{*}$ $(0.008)$ $-0.006$ $(0.007)$		(*******)	(0120)	(0.000)
$(0.024)$ $(0.014)$ $(0.005)$ Observations $14,715,945$ $14,715,945$ $14,715,945$ $\underline{Panel C. Men}$ US Employment Shock*1(exposed_h) $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ $1(exposed_h)$ $-0.028^{**}$ $(0.014)$ $-0.004^{*}$ $(0.008)$ $-0.006$ $(0.007)$	$\mathbb{1}(exposed_h)$	-0.044*	-0.024*	-0.011**
Panel C. Men US Employment Shock*1(exposed_h) $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ 1(exposed_h) $-0.028^{**}$ $(0.014)$ $-0.014^*$ $(0.008)$ $-0.006$ $(0.007)$		(0.024)	(0.014)	(0.005)
Panel C. Men US Employment Shock*1(exposed_h) $0.003$ $(0.168)$ $0.098$ $(0.125)$ $0.087$ $(0.072)$ 1(exposed_h) $-0.028^{**}$ $(0.014)$ $-0.014^*$ $(0.008)$ $-0.006$ $(0.007)$				
US Employment Shock*1(exposed_h) $0.003$ $0.098$ $0.087$ $(0.168)$ $(0.125)$ $(0.072)$ $1(exposed_h)$ $-0.028^{**}$ $-0.014^*$ $-0.006$ $(0.014)$ $(0.008)$ $(0.007)$	Observations	14,715,945	14,715,945	14,715,945
US Employment Shock*1(exposed_h) $0.003$ $0.098$ $0.087$ $(0.168)$ $(0.125)$ $(0.072)$ $1(exposed_h)$ $-0.028^{**}$ $-0.014^*$ $-0.006$ $(0.014)$ $(0.008)$ $(0.007)$				
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		0.000	0.000	
$     1(exposed_h) -0.028^{**} -0.014^* -0.006      (0.014) (0.008) (0.007)   $	US Employment Shock $\mathbb{I}(exposed_h)$			
$(0.014) \qquad (0.008) \qquad (0.007)$		(0.168)	(0.125)	(0.072)
$(0.014) \qquad (0.008) \qquad (0.007)$	1(exposed)	-0 028**	-0.01/1*	-0.006
	$\mathbb{I}(\exp(\operatorname{osecu}_h))$			
Observations 13,253,429 13,253,429 13,253,429		(0.014)	(0.008)	(0.007)
10,200,120 10,200,120 10,200,120	Observations	13 253 429	13 253 429	13 253 429
		10,200,120	10,200,120	10,200,120

Table C22: Cross-Sectional Employment Analysis 2000

This table examines whether labor supply behavior differs among households with and without unaffected US migrants in municipios facing different US shocks. We use cross-sectional data from the 2000 Census and define households exposed to US labor markets as those with either return migrants or with a household member living in the US. Columns (1) and (2) estimate the specification in Equation (8), including the main effect of the US Employment shock as a control, while column (3) estimates a more general specification with municipio fixed effects and thus omits the US Employment Shock main effect. Column (2) show the results including controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), share of employment in Mexico's non-tradable sector, and changes in homicide rates across municipios (and the municipio fixed effects in column (3) subsume all these controls). The negative coefficients for the interaction term in all columns of Panel A imply that the employment probability is higher in exposed households in municipios connected to larger US employment declines. Panels B and C show that the relationship is driven almost entirely by women. Results are qualitatively similar to those in columns (1) and (2), though a bit less precise, when controlling for state × exposure status fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

## C.8 Sample restriction

Tables C23 – C28 show results paralleling those in the main text but only limiting the sample to municipios whose citizens received at least 100 MCAS cards in 2006. For most outcomes, the point estimates are similar in magnitude and statistical significance to those in the main text, which further restrict the sample to municipios that had at least 5,000 residents in the year 2005 and had exposure  $\xi_s > 0.066$  (the 25th percentile). Notable exceptions are television ownership and primary school enrollment.

	Pop	ulation Gro	owth	Re	eturn Migra	tion		Emigratio	n
	(1)	2005-10			2005-10	(a)		2005-10	(0)
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
US Employment Shock	-1.684***	-0.684*	-0.330	-0.157**	-0.252***	-0.173**	0.222**	0.223**	0.259***
	(0.424)	(0.384)	(0.393)	(0.069)	(0.064)	(0.069)	(0.092)	(0.091)	(0.094)
Pre-shock Outcome		0.555***	0.544***		1.433***	1.414***		0.153***	0.144***
The shock outcome		(0.051)	(0.050)		(0.328)	(0.328)		(0.035)	(0.035)
New 287g Policy			0.243**			0.076***			0.015
Trew 201g I only			(0.107)			(0.016)			(0.010)
Employment Policy			0.005			-0.011			-0.005
			(0.064)			(0.011)			(0.015)
Trade Shock			-0.070			0.031***			0.041***
			(0.083)			(0.011)			(0.009)
Non-tradable share			-0.116**			-0.019***			0.005
of Employment			(0.051)			(0.007)			(0.010)
Homicide Rate			-18.044*			-4.703***			-7.880***
2005-10			(10.022)			(0.926)			(2.422)
State FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	$1,\!194$	$1,\!194$	1,194	$1,\!194$	1,194	$1,\!194$	1,192	1,192	1,192
R-squared	0.159	0.441	0.454	0.290	0.560	0.577	0.270	0.320	0.339

#### Table C23: Population Growth, Return Migration, and Emigration

This table examines the effect of changes in US labor demand on the 2005-10 population growth, return migration to, and emigration from each Mexican source municipio with more than 100 MCAS card issued. Note that outcome and pre-shock outcome variables are divided by exposure as in equation (7). We restrict attention to individuals age 15-64. Population growth is defined as the proportional change in population. Return migration is the number of individuals reporting living in the US 5 years prior to the relevant survey, divided by the municipio population in the survey year, while emigration is the number of household members who left for the US during the 5 years prior to the relevant survey, divided by the initial municipio population, measured using the roughly 10% long-form sample from the 2000 or 2010 Census (emigration information is not available in 2005). We use full-count tabulations from the 2000 or 2010 Mexican Census or 2005 Conteo to calculate population growth and return migration. All specifications in columns (1) to (9) use a GLS re-weighting procedure to address potential heteroskedasticity. The "Pre-shock Outcome" controls in columns (2), (5), and (8) are 2000-2005 population growth, 2000-2005 return migration, and 1995-2000 emigration, respectively. Columns (3), (6), and (9) additionally control for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios. All specifications control for Mexican state fixed effects, and standard errors clustered at the Mexican commuting zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

	0	$\triangle$ Sex Ratio (F/M) 2005-10	0.	ΔL	△ Less than Primary 2005-10	umary		$\triangle$ Primary 2005-10	~	-	△ Secondary 2005-10	IJ		$\triangle$ University 2005-10	ty	Reı	Household Remittances 2010	010
	(1)	(2)	(3)	(4)	(2)	(9)	(2)	(8)	(6)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)
US Employment Shock 0.632*** (0.215)	$0.632^{***}$ (0.215)	$0.530^{**}$ (0.197)	$0.474^{**}$ (0.213)	0.157 (0.098)	-0.036 (0.092)	-0.100 (0.101)	0.060 (0.115)	0.107 (0.121)	$\begin{array}{c} 0.114 \\ (0.131) \end{array}$	$-0.107^{**}$ (0.052)	-0.052 (0.059)	-0.055 (0.064)	-0.038 (0.042)	$0.084^{**}$ (0.039)	$0.094^{**}$ (0.044)	$0.555^{***}$ (0.110)	$0.462^{***}$ (0.096)	$0.495^{***}$ (0.107)
Pre-shock Outcome		$\begin{array}{r} -0.288^{***} & -0.287^{***} \\ (0.035) & (0.035) \end{array}$	$-0.287^{***}$ (0.035)		$0.513^{***}$ (0.022)	$0.506^{***}$ (0.023)		$0.595^{***}$ (0.079)	$\begin{array}{c} 0.614^{***} \\ (0.077) \end{array}$		$0.708^{***}$ (0.064)	$0.701^{***}$ (0.064)		$0.742^{***}$ (0.056)	$0.731^{***}$ (0.056)		$0.542^{***}$ (0.043)	$0.536^{**}$ (0.044)
New 287g Policy			-0.051 (0.044)			$-0.071^{***}$ (0.022)			$0.082^{***}$ (0.029)			-0.001 (0.015)			-0.005 (0.012)			0.003 (0.026)
Employment Policy			0.014 (0.032)			0.009 (0.016)			-0.013 (0.021)			-0.005 (0.010)			$\begin{array}{c} 0.001 \\ (0.008) \end{array}$			0.005 (0.017)
Trade Shock			$-0.083^{***}$ (0.025)			$-0.035^{**}$ (0.014)			$\begin{array}{c} 0.314^{***} \\ (0.038) \end{array}$			$-0.207^{***}$ (0.031)			-0.021 (0.025)			$0.038^{*}$ (0.022)
Non-tradable share of Employment			-0.005 (0.020)			0.012 (0.011)			$0.057^{***}$ (0.017)			$-0.022^{***}$ (0.008)			$-0.027^{***}$ (0.006)			0.019 (0.012)
Homicide Rate 2005-10			$9.665^{**}$ (3.817)			$2.995^{**}$ (1.515)			$2.916 \\ (1.780)$			0.921 (1.006)			-0.881 (0.864)			$-6.199^{**}$ (2.983)
State FE Observations R-squared	Yes 1,194 0.158	$\substack{\text{Yes}\\1,194\\0.281}$	$\substack{\text{Yes}\\1,194\\0.289}$	$\substack{\text{Yes}\\1,194\\0.202}$	Yes 1,194 0.727	Yes 1,194 0.730	Yes 1,194 0.263	Yes 1,194 0.461	Yes 1,194 0.498	$\substack{\text{Yes}\\1,194\\0.277}$	Yes 1,194 0.543	Yes 1,194 0.560	Yes 1,194 0.187	Yes 1,194 0.604	$\substack{\text{Yes}\\1,194\\0.606}$	Yes 1,194 0.146	Yes 1,194 0.397	Yes 1,194 0.403

Remittances
Receiving
Households
, and
ttainment
Sex Ratio, A
Sex B
C24:
Table (

outcome and pre-shock outcome variables are divided by exposure as in equation (7). We measure the sex ratio and educational attainment using the 2000 or 2010 Mexican Census or 2005 Inter-Censal Count. We calculate the share of households receiving remittances as the number of households reporting receiving income from relatives living abroad divided by the municipio's total number of households in the Census year, using the 2000 or 2010 Mexican Census. All specifications in columns (1) to (18) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1This table examines the effect of changes in US labor demand on the 2005-10 change in the female to male sex ratio for the working age population (15-64), the 2005-10 change in the share of the working age population in each education level for each Mexican source municipio, and the share of households receiving remittances in 2010. Note that

		EPOP 2004-09	1-09	$\Delta$ In	$\Delta \ln(\text{Hours}) 200$	2004-09	$\Delta \ln($	ln(Earnings) 2	2004-09	$\Delta \ln(E)$	$\Delta \ln(\text{EarnPerHour})$	c) 2004-09
	(1)	(2)	(3)	(4)	(2)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock	-0.047 (0.133)	-0.145 (0.140)	-0.125 (0.146)	-0.875 (1.114)	-1.103 (1.119)	-0.982 (1.287)	-1.294 (2.426)	-1.189 (2.433)	-1.644 (2.649)	-0.307 (2.036)	-0.265 (1.974)	-0.795 (2.216)
△ EPOP 1999-04		$-0.469^{***}$ (0.127)	$-0.453^{***}$ $(0.130)$									
∆ ln(Hours) 1999-04					-0.060 (0.066)	-0.090 (0.065)		$0.414^{***}$ (0.136)	$\begin{array}{c} 0.416^{***} \\ (0.138) \end{array}$		$0.538^{***}$ $(0.072)$	$0.531^{***}$ (0.073)
∆ ln(Earnings) 1999-04					$0.058^{***}$ (0.020)	$0.064^{***}$ (0.020)		$-0.138^{***}$ (0.049)	$-0.131^{***}$ (0.049)		$-0.189^{***}$ (0.033)	$-0.187^{**}$ (0.033)
New 287g Policy			-0.016 (0.046)			0.140 (0.412)			$1.132 \\ (0.774)$			$1.076^{*}$ (0.577)
Employment Policy			0.020 (0.027)			0.003 (0.168)			-0.792 (0.491)			$-0.816^{**}$ (0.379)
Trade Shock			$-0.453^{***}$ (0.075)			$-0.774^{**}$ (0.372)			-0.205 $(0.762)$			0.450 (0.518)
Non-tradable share of Employment			0.017 (0.029)			$0.554^{***}$ (0.200)			$1.063^{**}$ (0.448)			0.498 (0.333)
Homicide Rate 05-10			-1.331 (2.712)			-20.628 (19.554)			-34.998 (41.732)			-6.853 (30.580)
State FE Observations R-squared	$\substack{\text{Yes}\\1,194\\0.046}$	$\substack{\mathrm{Yes}\\1,194\\0.214}$	Yes 1,194 0.220	$\substack{\text{Yes}\\1,163\\0.194}$	$\mathop{\rm Yes}\limits_{0.134}$	$\substack{\text{Yes}\\1,160\\0.145}$	$\substack{\text{Yes}\\1,163\\0.102}$	Yes 1,160 0.140	$\begin{array}{c} \mathrm{Yes}\\ 1,160\\ 0.154 \end{array}$	$\substack{\mathrm{Yes}\\1,163\\0.077}$	$\substack{\mathrm{Yes}\\1,160\\0.203}$	$\substack{\text{Yes}\\1,160\\0.210}$

Table C25: Employment-to-nonulation Ratio and Earnings ner Hour (All Workers)

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio, earnings, hours worked, and earnings per hour in each municipio, using employment, earnings, and hours from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (12) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

	Δ	EPOP Wo	men		$\triangle$ EPOP M	len
		2004-09			2004-09	
	(1)	(2)	(3)	(4)	(5)	(6)
US Employment Shock	0.046	-0.163	-0.281*	-0.068	-0.097	0.023
	(0.093)	(0.139)	(0.145)	(0.181)	(0.173)	(0.191)
$\triangle$ EPOP		-0.496***	-0.492***		-0.384***	-0.390***
1999-04		(0.154)	(0.157)		(0.099)	(0.098)
New 287g Policy			-0.102***			0.061
			(0.039)			(0.075)
Employment Policy			-0.017			0.051
Employment Foncy			(0.025)			(0.038)
Trade Shock			-0.161			-0.819***
			(0.123)			(0.078)
Non-tradable share			0.023			0.034
of Employment			(0.037)			(0.036
Homicide Rate 05-10			0.521			-1.435
			(2.267)			(3.641)
State FE	Yes	Yes	Yes	Yes	Yes	Yes
Observations	$1,\!194$	$1,\!193$	$1,\!193$	$1,\!194$	$1,\!193$	$1,\!193$
R-squared	0.097	0.259	0.269	0.045	0.177	0.234

Table C26: Employment-to-population Ratio by Gender

This table examines the effects of declines in US labor demand on the 2004-2009 change in the employment-to-population ratio for women and men in each municipio, using employment from the 1999, 2004 and 2009 Mexican Economic Census and population from the 2000 and 2010 Mexican Census and the 2005 Conteo. Note that the outcome and pre-shock outcome variables are divided by exposure as in equation (7). We trim the bottom and top 1 percent of the earnings distribution. All specifications in columns (1) to (6) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p<0.01, ** p<0.05, * p<0.1

		$\triangle$ Computer 2005-10	er	>	△ Washing Machine 2005-10	achine	Z	△ Kehrigerator 2005-10	tor		$\triangle$ TV 2005-10	
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)
US Employment Shock -0.016 (0.105)	-0.016 (0.105)	0.075 (0.074)	0.045 (0.079)	0.238 (0.190)	$0.347^{*}$ (0.179)	0.245 (0.200)	-0.075 (0.197)	$0.104 \\ (0.176)$	0.017 (0.187)	0.010 (0.182)	0.203 (0.212)	0.127 (0.236)
∆ in Outcome 2005-00		$0.833^{***}$ (0.032)	$0.830^{***}$ (0.031)		$0.262^{***}$ (0.020)	$0.254^{***}$ (0.020)		$0.292^{***}$ (0.026)	$0.285^{***}$ (0.027)		$0.315^{***}$ (0.026)	$0.316^{**}$ (0.027)
New 287g Policy			-0.035 (0.025)			0.067 (0.047)			0.067 (0.044)			-0.007 (0.040)
Employment Policy			-0.007 (0.016)			$-0.093^{***}$ (0.032)			$-0.075^{**}$ (0.030)			-0.024 (0.029)
Trade Shock			$-0.093^{**}$ (0.038)			0.048 (0.064)			$0.097^{*}$ (0.056)			0.074 (0.048)
Non-tradable share of Employment			$-0.080^{**}$ (0.061)			0.037 (0.053)			$0.094^{*}$ (0.041)			0.061
Homicide Rate 05-10 of Employment			0.569 (1.352)			-1.248 (2.734)			1.577 (2.774)			$6.036^{**}$ (2.879)
State FE	$Y_{es}$	$Y_{es}^{1}$	$Y_{es}$	$Y_{es}$	${ m Yes}_{10.4}$	Yes 1 104	$Y_{es}$	${ m Yes}_{1.10.4}$	$Y_{es}$	${ m Yes}$	$Y_{es}$	$Y_{es}$
Cusei vauouis R-squared	0.262	0.856	0.857	0.414	0.540	0.544	0.342	0.517	0.521	0.189	0.376	0.379

Table C27: Appliance Ownership

and televisions). We calculate the change in the share households owning the relevant household durable using the 2000 or 2010 Mexican Computers, washing machines, refrigerators, in columns (1) to (12) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include Mexican State fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parenthese. *** p<0.01, ** p<0.05, * p<0.1This ta

	$\triangle A$	ttendanc	e Rate	Δ	Attendance	Rate	$\triangle$	Attendance	Rate
	(age	es 6-12) 20		, 0	es 13-15) 20		( )	es 16-18) 20	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
US Employment Shock	-0.015 (0.059)	-0.020 (0.059)	-0.021 (0.067)	-0.258 (0.160)	-0.232 (0.165)	-0.120 (0.180)	$-0.391^{**}$ (0.195)	-0.286 $(0.200)$	-0.252 (0.219)
$\triangle$ in Outcome 2005-00		-0.023 (0.022)	-0.024 (0.023)		$0.110^{***}$ (0.028)	$0.099^{***}$ (0.028		$\begin{array}{c} 0.142^{***} \\ (0.033) \end{array}$	$0.128^{***}$ (0.034)
New 287g Policy			0.001 (0.016)			$0.060 \\ (0.048)$			$\begin{array}{c} 0.040 \\ (0.053) \end{array}$
Employment Policy			-0.004 (0.012)			$0.006 \\ (0.035)$			-0.030 (0.035)
Trade Shock			$0.035^{***}$ (0.012)			$\begin{array}{c} 0.103^{***} \\ (0.025) \end{array}$			-0.009 (0.044)
Non-tradable share of Employment			0.000 (0.009)			-0.026 (0.022)			$-0.079^{***}$ (0.028)
Homicide Rate 05-10			$-1.779^{*}$ (0.947)			-9.337** (3.880)			-4.059 (4.776)
State FE Observations R-squared	Yes 1,194 0.057	Yes 1,194 0.058	Yes 1,194 0.063	Yes 1,194 0.187	Yes 1,194 0.203	Yes 1,194 0.213	Yes 1,194 0.191	Yes 1,194 0.219	Yes 1,194 0.226

#### Table C28: School Attendance

This table examines the effect of changes in US labor demand on the 2005-10 change in school attendance. We calculate the change in the share of the population in primary (age 6-12), secondary (13-15), and high-school (16-18) reporting having attended school using the 2000 or 2010 Mexican Census or 2005 Conteo. All specifications in columns (1) to (9) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include Mexican state fixed effects. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. *** p < 0.01, ** p < 0.05, * p < 0.1

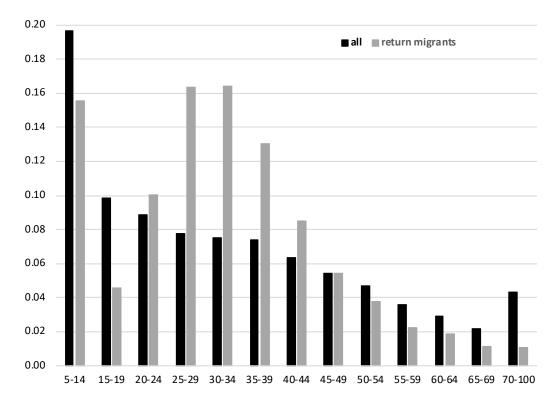
# C.9 Comparison of Return Migrants to Local Population

Table C29 and Figure C2 use 2010 Mexican Census data to compare return migrants to non-migrants.

Characteristics	All	Return Migrants 2005-2010
Female	51.3%	31.3%
Married	42.0%	57.9%
Education		
Less than primary	35.5%	25.8%
Primary	42.8%	54.5%
Secondary	14.5%	15.5%
University	7.2%	4.2%
Employment		
Employed	50.1%	62.9~%
Self-employed	27.7%	31.6%
Paid employee	69.1%	63.6%
Unpaid worker	3.2%	4.7%
Unemployed	2.4%	5.5%
Not in the labor force	47.5%	31.6%
Hourly wage (yr 2000 pesos)	19.64	18.90

#### Table C29: Descriptive Statistics, 2010 Census

Authors' calculations using data from the 2010 Mexican Census. Return migrants are defined as those living in the US in June 2005, five years prior to the Census. Hourly wages calculated as (monthly earnings / 4.33) / (weekly hours) and deflated to year 2000 pesos using the consumer price index (INPC) from INEGI. Average hourly wages omit the top and bottom 1 percent of observations. Note that, in comparison to the overall population, return migrants are disproportionately male, more likely to be married, more likely to have primary school education, and more likely to be in the labor force.



### Figure C2: 2005-2010 Return Migrants' Age Distribution

Authors' calculations using data from the 2010 Mexican Census. Return migrants are defined as those living in the US in June 2005, five years prior to the Census. Note that, in comparison to the overall population, return migrants are much more likely to be in the 25-39 age range and less likely to be under 20 and over 60.

#### C.10 Population Growth Decomposition

Table C30 provides a complete decomposition of the components of population growth, as examined in Table 1. In addition to return migration and emigration, this table considers net migration internal to Mexico, aging in and out of the working age population, and a residual component left over after removing all of these measured components. Columns (1), (4), and (7) exactly replicate the results in the respective columns in Table 1, and columns (10), (13), and (16) provide similar analysis for the additional components without any additional controls. The second column in each set of results controls for lagged changes in all of the dependent variables. Note that these specifications differ slightly from the second specifications shown in Table 1 because they control for multiple lagged changes rather than only the change in the dependent variable of a given regression. The final column in each set adds the further set of controls included in columns (3), (6), and (9) of Table 1.

The results reveal no statistically significant relationship between the US Employment shock and either net internal migration or population aging, which suggests that these components are not coincidentally related to the employment shock. Somewhat puzzlingly, the US Employment Shock is related to the residual portion of population growth. One possible explanation for this unexplained component is mismeasurement in migration. If, for example, some return migrants are coded as not having been in the US, perhaps due to concerns about having emigrated without legal authorization, it could lead to this pattern of results. Alternatively, whole-household emigrants are not captured by our emigration measure, so these emigrants may appear in the residual. Because we cannot say definitively what the source of the residual population growth is, we consider a range of values from 1 to 2 percentage points for the magnitude of the population growth implied by this analysis. The low end of the range reflects the implied change in population due to measured differences in net international migration, while the high end reflects the entire estimated change in population, including the residual component.

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Table

		Pol	Population Growth 2005-10	owth	Re	Return Migration 2005-10	ion		Emigration 2005-10	q	Net	Net Internal Migration 2005-10	gration	Pol in aı	Population aging in and out 2005-10	sing 5-10		Residual 2005-10	
Optimum         Jugar         <		(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)		(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)
0.333*         0.330*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         0.00*         <	US Employment Shock	$-2.070^{***}$ (0.573)	$-1.345^{***}$ (0.441)	$-1.098^{**}$ (0.440)	$-0.201^{***}$ (0.072)	$-0.210^{***}$ (0.059)	$-0.145^{**}$ (0.063)	$0.210^{*}$ (0.117)	$0.238^{**}$ (0.111)	$0.301^{***}$ (0.113)	$-0.477^{*}$ (0.274)	$0.022 \\ (0.217)$	$0.122 \\ (0.254)$	-0.512 (0.338)	$\begin{array}{c} 0.011 \\ (0.085) \end{array}$	$0.004 \\ (0.097)$	-0.505 (0.442)	$-0.984^{***}$ (0.376)	$-0.949^{**}$ (0.416)
- 3.40***         - 3.57***         - 1.07***         - 1.07***         - 1.07***         - 1.07***         - 0.811         - 1.02*         0.441         0.442         - 0.843         0.441         - 0.433         - 0.683         0.694)         - 0.693         - 0.683         0.694)         - 0.693         - 0.693         - 0.693         - 0.693         0.694)         - 0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693         0.693	Population Growth 2000-05		$0.363^{**}$ (0.142)	$0.356^{**}$ (0.142)		$-0.008^{*}$ (0.004)	-0.007* (0.004)		$-0.025^{***}$ (0.005)	$-0.024^{***}$ (0.005)		$0.206^{**}$ (0.092)	$0.212^{**}$ (0.092)		0.004 (0.008)	0.004 (0.008)		$0.222^{***}$ (0.077)	$0.219^{***}$ (0.077)
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Return Migration Rate 2000-05		$-3.442^{***}$ (0.292)			$1.072^{***}$ (0.407)	$1.056^{***}$ (0.394)		$1.173^{***} \\ (0.604)$	${1.142^{***}} \\ (0.600)$		-0.811 (0.426)	$-1.028^{*}$ (0.423)		0.444 (0.689)	0.442 (0.694)		-2.870***	-2.854**
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Emigration Rate 1995-00		$0.560^{***}$ (0.134)			$0.170^{***}$ (0.034)	$0.162^{***}$ (0.034)		$0.068^{**}$ (0.034)	$0.058^{*}$ (0.033)		$0.360^{***}$ (0.085)	$0.295^{***}$ (0.086)		$-0.068^{**}$ (0.034)	$-0.072^{**}$ (0.034)		$0.276^{**}$ (0.130)	$0.275^{**}$ (0.132)
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Net Migration Rate 2000-05		0.039 (0.026)	0.042 (0.026)		$0.010^{*}$ (0.005)	$0.011^{*}$ (0.006)		0.003 (0.002)	0.003 (0.003)		$\begin{array}{c} 0.914^{***} \\ (0.108) \end{array}$	$0.920^{***}$ (0.107)		$0.004 \\ (0.003)$	0.004 (0.003)		$-0.859^{***}$ (0.142)	$-0.864^{***}$ (0.139)
$ \begin{array}{[c] cliable large l$	Population aging in and out 2000-05		$0.562^{***}$ (0.088)	$0.548^{***}$ (0.091)		$0.020^{***}$ (0.005)	$0.016^{***}$ (0.005)		$0.052^{***}$ (0.008)	$0.051^{***}$ (0.008)		-0.041 (0.061)	-0.059 (0.061)		$0.924^{***}$ (0.012)	$0.924^{***}$ (0.012)		$-0.358^{***}$ (0.064)	$-0.355^{**}$ (0.066)
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	New 287g Policy			0.098 (0.108)			$0.063^{***}$ (0.016)			0.001 (0.025)			$0.149^{*}$ (0.077)			$\begin{array}{c} 0.000\\ (0.020) \end{array}$			-0.144 (0.111)
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Employment Policy			0.027 (0.011)			-0.004 (0.018)			0.014 (0.044)			-0.008 (0.015)			-0.005 (0.083)			0.059
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Trade Shock			-0.002 (0.010)			$0.026^{***}$ (0.015)			$0.038^{**}$			$0.224^{***}$ (0.012)			$-0.024^{**}$ (0.089)			-0.260***
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Non-tradable share of Employment			-0.047 (0.051)			$-0.012^{*}$ (0.007)			$0.019^{*}$ (0.011)			$\begin{array}{c} 0.012 \\ (0.037) \end{array}$			-0.000 (0.012)			-0.045 (0.053)
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	Homicide Rate 2005-10			$-20.781^{**}$ (9.274)			$-3.493^{***}$ (0.829)			$-6.451^{***}$ (1.599)			$-13.749^{***}$ (4.153)			-1.232 (1.104)			-1.874 (8.880)
	State FE Observations R-squared	Yes 866 0.163	Yes 866 0.588	Yes 866 0.595	Yes 866 0.332	Yes 866 0.599	Yes 866 0.617	Yes 866 0.271	Yes 866 0.394	Yes 866 0.411	Yes 866 0.104	Yes 866 0.706	Yes 866 0.714	Yes 866 0.224	Yes 866 0.978	Yes 866 0.978	Yes 866 0.088	Yes 866 0.602	Yes 866 0.608
roughly 10% long-form sample from the 2000 or 2010 Census (emigration information is not available in 2005). Internal net migration at the municipio level is calculated as	roughly 10% long-form sample from the 2000 or 2010 Census (emigration information is not available in 2005). Internal net migration at the municipio level is calculated as the difference between each municipio's immigrants and emigrants during the 5 years prior to the to the relevant survey, divided by the initial municipio population, measured using the roughly 10% long-form sample from the 2000 or 2010 Census (internal migration information is not available in 2005). We use full-count tabulations from the 2000 or	form sam veen each 10% long-	ple from t municipio form samı	the 2000 )'s immign ble from t	or 2010 C rants and the 2000 or	ensus (em emigrants r 2010 Cei	igration during tl nsus (inte	informat: he 5 yeau rnal mio	ion is not 's prior to ration inf	c available o the to th formation	e in 2005 he releval is not av	). Intern nt survey	al net mi _i , divided , 2005) V	by the in.	t the mu itial mun	nicipio le nicipio po	evel is compulation	alc 1, alc	ulated measur a 2000

2010 Mexican Census or 2005 Conteo to calculate population growth, return migration, and population aging in and out. All specifications use a GLS re-weighting procedure to address potential heteroskedasticity. The "Pre-shock Outcome" controls in columns (2), (5), (8), (11), (14) are 2000-2005 population growth, 2000-2005 return migration, 1995-2000 emigration, 1995-2000 net migration, and 2000-2005 population aging in and out, respectively. All specifications control for Mexican state fixed effects, and standard errors clustered at the Mexican commuting zone level are shown in parentheses. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses.

## C.11 Long-term changes in employment to population ratio

Table C31 provides additional analysis extending the follow-up window of the results in Table 4 into later periods following the onset of the Great Recession. The first column replicates the results shown in column (1) of Table 4. The second and third columns use the same regression specification but with differences in the dependent variable calculated through 2014 and 2019, respectively. The results in Panel B suggest that the increase in female labor supply due the loss of network-connected jobs lasted at least through 2019. For men (Panel C), there is no indication of a short-run or long-run change in employment rate.

	$\triangle$ EPOP	$\triangle$ EPOP	$\triangle$ EPOP
	2004-09	2004-14	2004-19
	(1)	(2)	(3)
Panel A. All			
US Employment Shock	-0.366**	-0.196	-0.712**
	[0.160]	[0.244]	[0.263]
	(0.186)	(0.269)	(0.362)
$\triangle$ EPOP	-0.511***	-0.490***	-0.294**
1999-04	(0.116)	(0.138)	(0.140)
Mean raw outcome	0.013	0.021	0.040
among less affected		0.004	0.010
Implied shock impact	0.007	0.004	0.013
Panel B. Women	0 500***	0 510**	1 101***
US Employment Shock	-0.563***	-0.519**	-1.131***
	[0.137]	[0.216]	[0.239]
	(0.177)	(0.233)	(0.291)
$\triangle$ EPOP	-0.577***	-0.577***	-0.429**
Women 1999-04	(0.159)	(0.156)	(0.168)
Women 1999-04	(0.109)	(0.100)	(0.108)
Mean raw outcome	0.015	0.025	0.042
among less affected	0.010	0.020	0.012
Implied shock impact	0.011	0.010	0.021
	0.011	01010	0.021
Panel C. Men			
US Employment Shock	-0.112	0.239	-0.275
r J	[0.250]	[0.391]	[0.354]
	(0.236)	(0.377)	(0.475)
		( )	( )
$\triangle$ EPOP	-0.408***	-0.377***	-0.167
Men 1999-04	(0.087)	(0.111)	(0.136)
	× ,	× ,	( )
Mean raw outcome	0.009	0.015	0.036
among less affected			
Implied shock impact	0.002	-0.004	0.005
State FE	Yes	Yes	Yes
Observations	865	865	865

Table C31: Employment-to-population Ratio: Extended Time Periods

This table examines the effects of declines in US labor demand on the change in the employment-to-population ratio in each municipio, using employment from the 1999, 2004, 2014, and 2019 Mexican Economic Census and population from the 2000, 2010, and 2020 Mexican Census and the 2005 Conteo to examine changes across different time horizons. Note that the outcome and pre-shock outcome variables are divided by exposure,  $\xi_s$ , as in equation (7). All specifications in columns (1) to (3) use a GLS re-weighting procedure to address potential heteroskedasticity. All specifications include controls for anti-immigrant employment legislation and new 287(g) agreements across US CZs, trade shocks across municipios (divided by 1,000,000), the share of employment in Mexico's non-tradable sector, changes in homicide rates across municipios, and Mexican state fixed effects. "Mean raw outcome among less affected" is the average of the dependent variable without dividing by exposure for municipios in the quartile with the smallest magnitude US employment shocks. "Implied shock impact" provides the predicted difference in the outcome (without dividing by exposure) for municipios with the 90-10 percentile difference in shock size (0.075) and average exposure (0.25). Standard errors clustered at the Mexican Commuting Zone level are shown in parentheses. Due to the shift-share structure of the US Employment Shock, we also present Borusyak et al. (2022) standard errors for this variable in square brackets. *** p<0.01, ** p<0.05, * p<0.1 based on standard errors in brackets when present.