Inside the Refrigerator: Immigration Enforcement and Chilling Effects in Medicaid Participation

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Abstract:

Economists have puzzled over why eligible individuals fail to enroll in social safety net programs. Recently, some have posited that "chilling effects" arising from an icy policy climate may explain low program take-up rates for immigrants, but such effects are inherently hard to measure. This paper investigates a concrete determinant of chilling, Federal immigration enforcement, and finds robust evidence that heightened enforcement reduces Medicaid participation among children of non-citizens. This is the case even when children are themselves citizens and face no eligibility barriers to Medicaid enrollment. Immigrants from countries with more undocumented U.S. residents and those living in cities with a high fraction of other immigrants are most sensitive to enforcement efforts. Up to seventy-five percent of the relative decline in non-citizen Medicaid participation around the time of welfare reform, which has been attributed to the chilling effects of the reform itself, is explained by a contemporaneous spike in immigration enforcement activity. The results imply that safety net participation is influenced not only by program design, but also by a broader set of seemingly unrelated policy choices.

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Given the widespread concern about moral hazard and crowd-out arising from social safety net programs, it is perhaps surprising that a high fraction of low-income individuals fail to participate in programs for which they are eligible. A Kaiser Family Foundation report estimates that 52 percent of eligible adults without private insurance took up Medicaid in 2002, for example (Davidoff et al, 2005). Take-up rates are particularly low for immigrants; just 30 percent of eligible non-citizen adults were enrolled in Medicaid in 2002, compared with 57 percent of citizens. (Davidoff et al, 2005).

A growing literature tries to understand why some refrain from using social assistance to which they are entitled. In the wake of the 1996 welfare reform and the associated decline in immigrant participation in public programs, some researchers posit that a generalized policy environment can affect program participation even for those who maintain eligibility. Such indirect effects are termed "chilling effects" because they arise from a generally icy policy climate rather than from direct eligibility changes. In practice, "chilling" has been treated as a residual that explains otherwise puzzling responses to changes in safety net programs. This paper investigates a previously unexplored determinant of chilling for immigrants, Federal immigration enforcement, to assess the extent to which the overall policy environment influences participation decisions in Medicaid.

As described below, the results suggest an economically and statistically significant relationship between the level of enforcement and participation in Medicaid by children of non-citizens, even when the children themselves are citizens. The results point to the importance of seemingly unrelated policy choices in determining participation in safety net programs.

I. Background

Economists interested in understanding take-up of public programs have emphasized the roles of stigma, information, and program design.¹ Though a full discussion of the take-up literature is beyond the scope of this paper, Remler and Glied (2003) and Currie (2004) offer reviews. Both conclude that the most consistent determinant of take-up is program design, including information provision, transactions costs, and the generosity of benefits.²

The take-up issue is more severe for immigrants. It is estimated that only 30 percent of eligible non-citizen adults are enrolled in Medicaid, for example. Immigrants may have particular difficulty obtaining information about programs, completing English application forms, and navigating the complex administrative system. Stigmatization of participation may be high for some immigrant groups (Bertrand et al, 2000). As discussed in Currie (2004), a sizable literature suggests that immigrant groups have higher eligibility for and lower take-up rates of public programs, and that assimilation facilitates take-up.

Until recently, the role of the broader policy climate in influencing program participation has received less attention. After welfare reform, however, there was a decline in program participation well beyond what would have been expected due to strict eligibility changes, especially for immigrants. As a result, some observers hypothesized that "chilling effects" due to the anti-immigrant language of the welfare reform bill may have

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¹ For example, Daponte, Sanders and Taylor (1999) find that providing information about Food Stamp eligibility to low-income households substantially increases participation rates, particularly fpr households with the most to gain from participation. Other studies explore how culture propagated through social networks could influence participation, perhaps due to stigma or information (Bertand et al., 2000, Borjas and Hilton, 1996, and Aizer and Currie, 2004.

² Despite its popularity as an explanation, there has been little empirical work successfully isolating the effect of stigma on program take-up.

discouraged immigrant participation in public programs for which they remained eligible.³

Though the existence of "chilling" arising from an icy policy climate is plausible, fear and informal dissuasion are difficult to observe. Analysts typically assume that otherwise unexplained declines in participation or take-up of non-citizens are due to chilling effects. Mazzolari (2004), for example, accounts for a wide range of economic and demographic factors and finds that non-citizen immigrants have an unexplained decline in take-up of several safety net programs of 3-4 percentage points following welfare reform. She attributes this excess decline to chilling.

Other literature exploits variation in state generosity towards immigrants following reform. The PRWORA bill removed Federal support for post-enactment immigrants (those arriving after August 1996) for the first five years of residence; states have the option to use their own funds to support this group. Borjas (2003) reports that non-citizen Medicaid participation fell more in less generous states. Noting that most non-citizens in the sample had arrived before 1996 and therefore maintained eligibility for Medicaid, Borjas surmises that declines in participation stemmed from the "chilling effects" of welfare reform. In contrast, Kaushal and Kaestner (2005) do not find differences in new immigrant participation in more and less generous states. However, they also interpret their results as evidence of "chilling effects," in this case arising from the icy national policy environment.

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³ The 1996 PRWORA welfare reform bill included a number of provisions that were targeted towards immigrants. Immigrant eligibility for public means-tested programs was restricted, even for legal noncitizens. For Medicaid, the law banned the use of federal funds for most post-enactment immigrants (those arriving after August 1996) for the first five years after arrival. States had the option to use their own funds to provide Medicaid to this group and about half of them chose to do so. The law also allowed states to ban legal pre-enactment non-citizen immigrants from participating in Medicaid, though almost all continued offering Medicaid to pre-enactment immigrants. In addition, the reform made it harder for states to use their own funds to provide benefits to undocumented immigrants. Welfare reform also restricted immigrant eligibility for food stamps, Supplemental Security Income, and cash welfare in ways that differed across states. Exceptions to immigrant restrictions were made for recently arrived refugees, Cuban/Haitian entrants, and some other groups.

In sum, several analyses have found that program participation decisions respond to policy changes in ways that extend beyond what would be expected based on the strict eligibility changes. The unexplained changes in participation decisions have been attributed to chilling. An Urban Institute report on the subject concludes:

"Because comparatively few legal immigrants were ineligible for public benefits as of December 1997, it appears that the steeper declines in noncitizens' than citizens' use of welfare, food stamps, and Medicaid owe more to the "chilling effect" of welfare reform and other policy changes than they do to actual eligibility changes." (Fix and Passel, 1999).

This paper takes a different approach by considering chilling induced by Federal enforcement of immigration laws. The Illegal Immigrant Reform and Immigrant Responsibility Act of 1996 and other budget changes in the 1990s substantially increased funding for border control and internal enforcement activities. As a result, enforcement levels increased. The number of internal apprehensions (referred to below by the Immigration and Naturalization Service term "deportable aliens located") went from 70,000 in 1995 to 123,000 in 1997, for example.

There are good reasons to believe that Immigration and Naturalization Service (INS) policy could affect program participation. For example, following Proposition 187's passage in 1994 in California, the Department of Health Services developed a program with the Immigration and Naturalization Service to request repayment of Medicaid benefits for non-citizen immigrants upon re-entry into the United States after a trip abroad. Other anecdotes suggest that applicants for citizenship were occasionally asked to reimburse the government for previously used benefits, though this was not official policy.

For undocumented immigrants seeking health insurance for their children, fear of government authority is a natural concern. Loue, Cooper and Lloyd (2005) interview 157 women in San Diego in 1999-2001 and find that roughly a quarter of immigrants arriving

after 1996 and a quarter of undocumented immigrants had heard that they could not obtain medical care due to immigration status. Similar proportions said they were somewhat or very afraid to obtain medical care for themselves or a family member.

Program design and the general policy climate have the potential to exacerbate or ameliorate fears of undocumented immigrants. For instance, application forms for means-tested typically require or request Social Security numbers for every member of the household, even if only children are applying for benefits.⁴ Of six welfare sites studied in a 2003 report for the Department of Health and Human Services, only one explicitly states that Social Security numbers will be used only to verify income and will not be shared with the Immigration and Naturalization Service. On the other hand, applications at two sites explicitly state that information will be shared with the INS and that the INS response could affect benefit levels or lead to an investigation.⁵

INS policy could influence the program participation decisions even for legal permanent residents. For example, the welfare reform bill reiterated a long-standing doctrine that immigrants deemed a "public charge" could be deported or denied future citizenship. Though "public charge" deportations have rarely been implemented in the post-war period, the term was not defined in the legislation. It was not until late 1997 that a clarification was made indicating that occasional use of safety net services would not be grounds for deportation or denial of citizenship. Nevertheless, even after that date there were reports of immigrants being told that participation in public programs could jeopardize their immigration status (Schlosberg and Wiley, 1998).

To investigate the interactions between program participation and enforcement of immigration law, this paper exploits spatial and temporal variation in enforcement action between 1993 and 2002. The increase in immigration enforcement in the 1990s varied

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⁴ Recently some states have been removing requests for household social security numbers on application forms in an effort to increase Medicaid and State Children's Health Insurance Program participation among children of undocumented immigrants (Holcolmb et al., 2003)

⁵Holcomb et al., 2003.

substantially across the 33 INS administrative districts and across country-of-origin groups.

To distinguish the impact of enforcement from potential confounding factors, it is helpful to understand what drives variation in enforcement within a district over time. The bureaucracy of the INS is generally perceived to leave a large amount of discretion to district managers. Many observers lament the lack of centralized decision making and the absence of communication between districts (see Martin, 2000, for example). Therefore, idiosyncratic preferences of district managers combined with aggregate budget fluctuations are likely important determinants of changes in the degree of immigration enforcement over time within districts.

There could also be demographic changes contributing to the level of enforcement. For example, new waves of undocumented immigrants might trigger an increase in enforcement activity. To mitigate the potential bias from this source, non-citizen children whose mothers arrived less than five years prior to the survey year are excluded from most analyses. Some analyses also exclude citizen children born to newly arrived mothers. The identifying assumption is that within-area changes in enforcement are not correlated with unobservable factors that could differentially affect program participation decisions of children of non-citizens. Below I find little suggestion that enforcement activity is correlated with *observable* characteristics of non-citizen families in the sample, a fact which lends support to the identification strategy.

The analysis described below aims to consider a novel determinant of "chilling" and to shed light on how policies unrelated to program design could affect program participation. There has been little previous work examining the link between enforcement and program participation.⁶

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⁶ One exception is unpublished work by Vargas (2010) who explores the effect of fear of deportation on WIC and SCHIP participation for immigrants in mixed status families. Previous research has looked at the effect of enforcement on labor market outcomes. Bansak (2005) finds a negative effect of employer sanctions on wages of likely illegal immigrants in the 1980s, for example. Orrenius and Zavodny (2009) report adverse labor market consequences for Latin American immigrants post-2001 which they attribute to increased enforcement.

II. Data.

Information on Medicaid participation comes from the March Annual Demographic Supplements to the Current Population Survey (CPS), a survey run by the U.S. Census Bureau which aims to be nationally representative of households in the United States. The CPS asks whether each individual in the household was covered by Medicaid in the previous calendar year and is among the most commonly used data sets in studies of Medicaid participation. Citizenship status and country of origin of each household member is available starting in the 1994 survey. The survey contains a number of other demographic and economic indicators as well.

I pool the March surveys for the years 1994-2003 to generate the sample, which covers the reference years 1993-2002. My sample is limited to children under 18 years of age who can be matched to a mother within the household. I also exclude children directly targeted by the provisions of the 1996 PWRORA bill: non-citizen children whose mothers arrived less than five years prior to the survey. Another advantage to excluding this group is that it mitigates bias coming from new immigrant inflows determining enforcement activity. The primary analysis is based on a low-SES sample, which is limited to children below 200 percent of the poverty line whose mothers lack a college degree.

I assign children's status based on their mother's country of origin and citizenship status, under the assumption that mothers are likely to make decisions about Medicaid enrollment for the family.⁸ Unfortunately, the immigration status of non-citizens (i.e. whether they are documented or undocumented immigrants) is not observable in the CPS.

Under-reporting of program participation is an important limitation of these data. Meyer, Mok, and Sullivan (2009) find substantial under-reporting of public benefit receipt

⁷ Undocumented immigrants are likely to be undercounted in the Current Population Survey.

⁸ Alternative methods of assigning child's status are also explored.

compared to administrative records in 5 major surveys, including the Current Population Survey. The Meyer *et al.* study does not examine Medicaid participation, but finds reporting rates of only 50-70 percent for AFDC/TANF in the CPS. Medicaid misreporting may be a more serious problem because state Medicaid programs have multiple names and Medicaid may lack the salience of cash welfare for participants. Of particular concern is the potential that under-reporting behavior is responsive to enforcement. Unfortunately, it is very difficult to evaluate the degree to which selective misreporting drives the results, and I treat the survey responses as reflecting actual participation.

As is common in the literature, I use the data available in the CPS to impute each child's Medicaid eligibility. This imputation includes measurement error. For example, individuals with high levels of medical expenses may qualify for Medicaid but appear ineligible, whereas individuals with high levels of assets may be disqualified but appear eligible. I use two alternative measures of eligibility. The first imputes AFDC/TANF eligibility in addition to "expansion eligibility", where "expansion eligibility" includes children with family income low enough to qualify for Medicaid even if their family does not qualify for AFDC/TANF. Because AFDC/TANF eligibility is difficult to measure, a second definition of Medicaid eligibility relies on expansion eligibility only. Both measures of eligibility are imperfect, and analyses that examine take-up (rather than overall participation) should be interpreted with some caution.

The analysis also requires information on state welfare policy and immigration attitudes. For welfare policy, I rely on detailed information provided by Zimmerman and Tumlin (1999) on state welfare policies related to immigrants following welfare reform. I also use the American National Election Study (ANES) to calculate state-level measures of attitudes towards immigration. The ANES asks the respondent whether he or she would like to see immigration increased, unchanged, or decreased in the years 1992, 1994, 1996, 1998, 2000, and 2004. The answers to this question are collapsed into a binary variable

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⁹ Many thanks to Lara Shore-Sheppard for sharing the imputation algorithm and eligibility rules.

¹⁰ The National Election Studies (www.electionstudies.org). THE 2004 NATIONAL ELECTION STUDY [dataset]. Ann Arbor, MI: University of Michigan, Center for Political Studies [producer and distributor].

indicating whether or not the respondent would like to see immigration reduced and then aggregated to the state level. For states without responses, the average of the Census region is used. Unfortunately, small sample sizes in the ANES mean that this variable does not offer much predictive power.

Immigration enforcement data was obtained from the Department of Homeland Security via a 2009 Freedom of Information Act request. The dataset covers fiscal years 1992 to 2003 and consists of counts of Immigration and Naturalization Services "deportable aliens located" as the result of internal investigations, by INS internal district, country of origin, and fiscal year. "Deportable aliens located" is the INS term for apprehensions. Because some cells are suppressed due to confidentiality concerns, these data are supplemented with published reports in the INS Statistical Yearbooks listing deportable aliens located by INS district and fiscal year.

I aggregate the 33 INS districts into 25 "clusters" of states which map into CPS geography.¹¹ The level of enforcement in a fiscal year is summarized by the number of deportable aliens located divided by the estimated number of non-citizens in 1995.¹² The log average enforcement over a two year period including the year prior to and year of the Medicaid decision is the indicator of enforcement activity.

The sample excludes the small number of children directly targeted by the welfare reform bill: non-citizen children whose mothers arrived within five years of the survey date. The bulk of the analysis focuses on a low socioeconomic status sample – defined as children in families below 200 percent of the Federal poverty line in the reference year and whose mothers have less than a college degree.

Table 1 shows the key summary statistics for the children in the low-SES sample and the full sample. Medicaid participation is highest for children of non-citizens. Such children

¹² I estimate the number of non-citizens using IPUMS Census data for 1990 and 2000. The average of these two numbers is the estimated population for 1995.

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¹¹ Clusters are usually a single state or a group of states. The one exception is that the New York metropolitan area within New York state is an independent INS district and its own cluster. INS districts typically follow county lines and are often states or groups of states.

have less educated mothers but are less likely to live in single parent families. Children of non-citizens are also more likely to be income-eligible for Medicaid and to lack health insurance.

Figure 1 shows trends in enforcement over time. There is a sharp increase in enforcement in the mid-1990s, presumably due to the sharply increasing INS budget and manpower.¹³ The pattern of non-citizen Medicaid participation inversely mirrors the trends in enforcement over time, with a sharp drop in the 1997-1999 survey years followed by a rebound. Of course, any number of factors could be affecting both enforcement and participation, so national trends in both of these variables will be held constant in the analysis.

Figure 2 reports one measure of enforcement activity - the number of deportable aliens located (normalized by the estimated number of non-citizens in 1995) - by fiscal year for 7 of the 25 INS clusters in the data. Some areas, such as Texas, experienced sharp increases in enforcement activity while others, such as California, saw more modest changes.

III. Methodology and Results

a. Enforcement and Non-Citizen Medicaid Participation

The analysis examines the effect of immigration enforcement on Medicaid participation by children of non-citizens. For an overview of the data, I start by considering a sample of children of non-citizens only. The preliminary linear probability model is:

$$Medicaid_{ict} = \beta_0 + \beta_1 enforce_{ct} + \theta_c * year + \delta_t + \mu_{ict}$$

¹³ Full-time equivalent staffing for internal immigration enforcement jumped from 1746 in fiscal year 1995 to 2513 in fiscal year 1998. The overall enforcement budget increased from 2.1 billion to 3.4 billion over the same time period, and the share of those funds spent on border control declined from 64 to 56 percent, leaving additional resources for internal enforcement and investigations. (Source: "Immigration Enforcement Spending Since IRCA," Migration Policy Institute Fact Sheet, November 2005.)

where *enforce* refers to INS enforcement activity in cluster c relevant for participation year t, θ_c interacted with *year* controls for a cluster-specific linear time trend, and time fixed effects λ_t control for shocks that affect all non-citizens nationally. Standard errors are clustered by INS cluster to account for common shocks in a given local area.

Table 2 shows the results for the low-SES sample and the overall sample of children of non-citizens. One log-point increase in enforcement activity in one's local area reduces Medicaid participation by 8.7 percentage points for low-SES children and 4.9 percentage points for all children. It is also evident from Table 2 that there is no comparable effect on children of non-citizens, suggesting that the results for the non-citizen sample are not generated by factors discouraging Medicaid participation more generally. Furthermore, there are no comparable effects if one considers the *lead* in enforcement, where the lead is defined as the average of the survey year (the year following the reference year) and the subsequent year. These results suggest that enforcement reduces Medicaid participation for children of non-citizens.

b. Full Analysis of Participation

To improve statistical power and to more fully account for local shocks, the bulk of the analysis combines non-citizens and citizens and looks for a *differential* response to enforcement activity. The preferred specification is a linear probability model:

$$Medicaid_{icsgt} = \beta_0 + \beta_1 enforce_{ct} * noncit_i + \beta_2 enforce_{ct} + \Omega_{sgt} * noncit_i + \lambda_t * noncit_i + \psi_{gt} + X_iB_3 + \mu_{icsgt}$$

where *enforce* refers to INS enforcement activity in cluster c relevant for participation year t, $noncit_i$ indicates that the mother of child i is a non-citizen. Controls account for state-group-citizen fixed effects $\Omega_{sgt}*noncit_i$ to capture permanent state differences facing children of non-citizens of a particular country-of-origin group, ¹⁴ year dummies δ_t interacted with $noncit_i$ to account for annual changes in non-citizen participation nationally. Time flexible controls for country of origin group ψ_{gt} and other control

¹⁴ The New York City metropolitan area and the remainder of New York are treated as separate "states" because they are located within separate INS clusters.

variables are also included in most specifications. Standard errors are clustered on INS cluster to account for common shocks. In this specification, the key coefficient β_1 represents the effect of enforcement on children of non-citizens over and above the effect of enforcement on other children.

Panel A of Table 3 shows the main results for the low-SES sample with different sets of controls. The preferred specification (second column) shows that one log point increase in enforcement efforts reduces Medicaid participation by children of non-citizens by 9.9 percentage points. One can also restrict to citizen children, children whose mother's arrived more than five years ago, or both. Results are largely comparable for these groups. That is, even for children born in the U.S. to long-standing residents, enforcement influences the Medicaid participation decision.

Panel B of Table 3 shows comparable results for the full sample. Point estimates are roughly half the size for this group because few high-SES children participate in the Medicaid program. Nevertheless, even in the full sample the preferred specification (second column) shows a statistically significant reduction in Medicaid participation for children of non-citizens.

c. Does Enforcement Predict Eligibility or Other Characteristics?

Table 4 explores the implications of enforcement for Medicaid eligibility. We might be worried about such a relationship if enforcement changes coincided with state eligibility expansions that disproportionately benefited non-citizens, or if economic conditions changed such that fewer non-citizens were eligible. I impute eligibility in two ways, as described in the data section. The first incorporates the AFDC/TANF pathway and eligibility arising due to Medicaid expansions.¹⁶ The second ignores AFDC/TANF

¹⁵ Recall that non-citizen children whose mother's arrived less than five years ago are potentially directly affected by welfare reform and are therefore excluded from all analyses. The results are not substantively changed if this group is included.

¹⁶ To impute eligibility for TANF after 1996, I use AFDC rules in place in 1996. For subsequent years, states were required to offer Medicaid to those children who would have been eligible under AFDC rules.

eligibility pathway because that pathway is difficult to measure. The first column of Table 4 replicates the preferred participation results. Immigration enforcement is not predictive of Medicaid eligibility for the low-SES sample or the overall sample (second and third columns of Table 4).

Table 5 examines the effect of immigration on take-up of Medicaid – that is, participation conditional on eligibility. The effects of enforcement on take-up are of similar magnitude to the effects on participation for the low-SES sample. For the high-SES sample, the effect on take-up is larger than the effect of participation. This is not surprising given that less than a third of the overall sample is estimated to be eligible for Medicaid. In sum, it appears that enforcement is not correlated with income eligibility and discourages take-up conditional on eligibility for children of non-citizens.

In Appendix Table 1, I explore whether enforcement is predictive of other observable factors that might influence participation. These include family poverty status, mother's marital status, mother's education, mother's labor supply, child's age, and mother's time of arrival. There is no statistically significant relationship between enforcement and any of these factors. This fact lends support to the notion that Medicaid participation is influenced by enforcement rather than by contemporaneous economic or demographic changes across areas that disproportionately affect non-citizens.

d. Is Chilling National or Local?

In Table 6, I explore alternative dimensions of enforcement for the foreign born Low-SES sample. Column I repeats the preferred analysis using enforcement at the INS cluster level. Column II instead considers enforcement targeted at one's country-oforigin group at the national level, and finds that it is not predictive of Medicaid participation.¹⁷ Similarly, group-specific enforcement in one's cluster has a smaller

States also have work requirements and other policies that shape eligibility for TANF; these are not fully captured by my imputation algorithm.

¹⁷ Because the regressions control for non-citizen*year effects, this is the effect of enforcement targeted towards one's group over and above aggregate national changes in enforcement.

effect than overall enforcement in one's local area and has a statistically insignificant effect on participation.¹⁸ It appears that aggregate local enforcement is the most important determinant of participation, though some caution is warranted due to measurement error in the local group-specific variable.

Unfortunately, measures of enforcement are unavailable at geography smaller than the INS district. However, it is possible that metropolitan areas with many non-citizens experience a disproportionate share of district enforcement per non-citizen. Furthermore, even if enforcement is proportional to the number of non-citizens across cities within a district, residents of areas with many non-citizens might be more aware of enforcement policy.

To explore effects in high- and low-non-citizen areas, I find the fraction non-citizen of the total population for each of the 201 metropolitan areas in the sample. The median level of fraction non-citizen is computed for each country-of-origin group in the sample, and for each group the sample is split into those above and below the median. As is evident in column V of Table 6, there is an insignificant effect of cluster-level enforcement for non-citizens residing in areas with few other non-citizens, though the coefficient is of similar magnitude to the full sample. Columns VI through VIII suggest no statistically detectible pattern relating enforcement and Medicaid participation.

In areas with many non-citizens, on the other hand, the effect of aggregate enforcement at the cluster level is quite pronounced (see the columns IX of Table 6). There is also a statistically significant effect of local group-specific enforcement in column XI. In the horse race in column XII, aggregate cluster-level enforcement appears to be more important that group-specific local enforcement, but this may be due to measurement

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¹⁸ The relatively weak results for group-cluster enforcement may stem from measurement error. Local group-specific enforcement was suppressed for small cells in the enforcement data.

¹⁹ I calculate exposure to non-citizens by averaging the fraction non-citizen in the metropolitan area in the 1990 Census and fraction non-citizen in the 2000 Census. The median is constructed separately for each country of origin group because groups that cluster in non-citizen areas may also respond differently to enforcement for other reasons. I combine those at and above the median into a single group; the results are not sensitive to this choice. Results are also quite similar if exposure to non-citizens is replaced with exposure to same-group members.

error in the latter variable. The more substantial impact of enforcement in non-citizen enclaves may arise because enforcement per non-citizen is disproportionately located in these areas, because immigrants have more access to information about enforcement actions, because immigrant social networks are more likely to include someone affected, or some combination of these factors.

I also separate immigrants into groups with more and less than 20 percent of non-citizens estimated to be undocumented. The effect of enforcement is close to zero and insignificant for groups in which most immigrants are documented. The effect of enforcement is concentrated among groups with many undocumented immigrants (results not shown).

e. Citizenship

One potential threat to identification is that individuals have some ability to decide whether to become citizens, and they may pursue citizenship if the policy climate is less favorable towards non-citizens. Indeed, rates of citizenship increased substantially over the sample period.²⁰ Van Hook (2003) argues that the changing composition of citizenship may explain up to half of the decline in welfare participation following welfare reform. To investigate the possibility of endogenous citizenship, I first examine whether the probability that a child's mother is a citizen appears to respond to enforcement. I do not find evidence that this is the case, perhaps because it usually takes five years of legal residence plus a year or more of processing time to become a citizen.²¹

To further investigate this question, I instrument for mother's citizenship using her country of origin.²² In this framework one can control for state-group fixed effects but

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²⁰ Van Hook (2003) notes that the number of naturalizations was 240,000 in 1992 and peaked in 1996 at over one million.

²¹ Results not shown. In a regression with mother non-citizen on the left hand side and including state-group fixed effects and education controls, the coefficient on enforcement is 0.011 with a standard error of 0.009; in other words, enforcement has an insignificant and wrong-signed coefficient. High application fees may further deter would-be citizens. Immigrants married to citizens and those serving in the military have shorter residency requirements.

²² The IV regressions include state-group fixed effects rather than state-group-citizen fixed effects.

not state-group-citizen fixed effects. The OLS analysis specification using the revised specification with state-group instead of state-group-citizen fixed effects yields a smaller but statistically significant differential effect of enforcement on children of non-citizens. The instrumented coefficients are larger than the OLS estimates and similar to the baseline effects reported on Table 7, suggesting that endogenous citizenship is not driving the results. This test does not rule out the possibility of selective return migration or survey non-response by those fearing enforcement, but does suggest the relationship between enforcement and Medicaid participation does not arise because of selective maternal entry into citizenship.

f. Insurance Status and Program Participation

Table 8 presents information on how enforcement affects insurance status. The effect of enforcement on public health insurance is almost identical to the effect of Medicaid. This suggests that immigrants deterred from Medicaid due to enforcement are not enrolling in alternative public health insurance programs.²³ Private health insurance increases slightly in response to enforcement for the low-SES sample. The point estimates from Panel A imply that a 10 percentage point increase in Medicaid participation (due to absence of enforcement) crowds out 1.7 percentage points of private insurance for the low-SES sample. However, the crowd-out "ratio" (the change in private insurance divided by the change in Medicaid) would have standard errors too large to generate a precise crowd-out estimate.

Table 8 also suggests that the reduction in a child having any health insurance (6 percentage points due to a log-point increase in enforcement) is only about 60 percent of the reduction in Medicaid, implying 4 percentage points of "crowd-out" associated with a 10 percentage point change in Medicaid. The discrepancy between the change in private insurance and the change in any insurance stems from the fact that about 8 percent of low-SES children participate both in Medicaid and private insurance over the course of a

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²³ Alternative public programs could include idiosyncratic state programs, Indian Health Service programs, military insurance programs, etc.

year.²⁴ Thus, the point estimates suggest that of every 100 children discouraged from Medicaid participation due to enforcement, 17 enroll in private insurance that they would not have otherwise had, and another 23 rely exclusively on private insurance that they would have had for part or all of the year.²⁵ These are not precisely estimated numbers, however. Panel B presents results for the full sample, in which the implied crowd-rates are smaller.

Table 8 also presents the estimated effect of enforcement on other poverty programs. The impact on participation in public assistance (AFDC/TANF) is small and statistically insignificant. I also find negligible impacts of enforcement on the probability that a family receives Disability Income or Supplemental Security Income (results not shown). On the other hand, there is suggestive evidence that Food Stamp participation has a similar response to enforcement as Medicaid, though standard errors are large.

One explanation for the greater sensitivity of non-cash benefits to enforcement is that undocumented immigrants may be reluctant to apply for cash benefits even in the absence of enforcement. As noted by a 2003 Health and Human Services report, health benefits typically have the most streamlined application process (often possible to complete by mail or internet) and are least likely to require an in-person interview or fingerprinting. Food stamp applications are often integrated with the cash assistance application and tend to be complicated, but some states have integrated a Food Stamp screen into the Medicaid/SCHIP determination process and some states have stand-alone food stamp application locations (Holcomb *et al.*, 2003). An in-depth analysis of the effect of enforcement across programs is beyond the scope of this paper, but the differences suggest that enforcement can interact with program design to influence participation.

²⁴ About 21 percent of low-SES children with private health insurance during some point of the year also have public health insurance during some point in the year. About 17 percent of low-SES children with public health insurance also have private insurance. The CPS does not offer information about whether these sources of insurance are concurrent.

²⁵ The different implied crowd-rates highlights the importance of considering within-year insurance transitions. See Buchmueller and Shore-Sheppard (2010).

g. State Policy Climate

The chilling literature has emphasized state policy generosity towards immigrants. It is important to account for state policy changes around the time of welfare reform in the analysis of the effect of enforcement on Medicaid participation. Researchers have used various criteria to categorize as a state as generous.

For example, Borjas (2003) considers a state "generous" if it offered food assistance or SSI to pre-enactment immigrants or offered any of four major programs (TANF, Medicaid, food assistance, or SSI) to post-enactment immigrants.²⁶ This definition includes the six largest immigrant states; 89 percent of children of non-citizens in my sample live in a generous state according to the Borjas definition.²⁷

Kaushal and Kaestner (2005) offer a simpler definition, describing a state as "generous" if it offered TANF or Medicaid to post-enactment immigrants. Under this definition 56-57 percent of children of non-citizens live in generous states. Among the six largest immigrant states, only California and Illinois are considered generous.

As a third alternative, I consider a measure of state generosity based on rankings of the availability of state safety net programs to immigrants in a 1999 Urban Institute report. I describe states as generous if Zimmerman and Tumlin (1999) categorize immigrant safety net programs in the state as most available or somewhat available. All of the major immigrant states except Texas are included as generous; 72 percent of children of non-citizens live in generous states according to the Zimmerman and Tumlin definition. For all three measures of generosity, the state is labeled as generous or not generous after welfare reform and the generosity variable equals zero for all states prior to welfare reform.

²⁶ Post-enactment immigrants are those arriving after welfare reform in August 1996.

²⁷ The six states with the highest numbers of immigrants are California, Florida, Illinois, New York, New Jersey, and Texas.

I also incorporate a measure of population anti-immigrant sentiment from the ANES, as described in the data section. The variable can be interpreted as the fraction of all residents of a state who would prefer to see the level of immigration reduced.

Table 9 shows the effect of state climate using these measures. The first column shows no detectible effect of local immigration attitudes on Medicaid participation, though it is possible that measurement error generates a lack of statistical power. Of the three measures of state welfare generosity, only the Zimmerman and Tumlin definition shows a nearly significant relationship with participation.

In the second four columns of Table 9, state characteristics are included in the analysis of enforcement. Only one of the four weakens the effect of enforcement (the last column of Table 9). Even including Zimmerman and Tumlin generosity as a control, enforcement is a significant predictor of Medicaid participation for children of non-citizens.

h. Robustness

Appendix Table 2 presents the results of sensitivity analysis. The preferred specification is replicated in the first column. The second column shows the results using a linear rather than logged measure of enforcement. The results suggest that a one percentage point increase in enforcement (e.g. increasing from one arrest per 100 non-citizens to two arrests per 100 non-citizens) reduces Medicaid participation by 5.8 percentage points. This effect is of the same order of magnitude as that implied by the log specification evaluated at the sample mean.

The third column of Appendix 2 restricts the sample to mothers arriving in the U.S. prior to 1992. This sample restriction mitigates concerns about the correlation of enforcement activity with unobservable characteristics of new migrants. Finally, the fourth column incorporates state-specific linear time trends. I also try dropping each of the six large immigrant states on at a time (results not shown). The results are robust to these alternative specifications.

Appendix Table 3 explores alternative definitions of citizenship. The bulk of the analysis uses the mother's citizenship status to predict Medicaid participation. Results are similar if the mother's spouse is a non-citizen, if either parent is a non-citizen, or if both parents are non-citizens. The final two columns of Appendix Table 3 show that having a non-citizen spouse makes a citizen mother much more responsive to enforcement but has a relatively minor effect on a non-citizen mother. In sum, families are responsive to enforcement when either parent is a non-citizen.

i. Magnitude of the Effects

To gauge the magnitude of the effects, I use the estimated model to predict what would have happened to Medicaid participation among children of non-citizens if enforcement levels maintained their levels early in the sample period – specifically, the average of 1993 and 1994 levels. The results suggest that participation would have fallen from 46.5 percent in survey year 1995 to 45.4 percent in survey year 2000, a drop of 1.1 percentage points, had enforcement stayed constant at the 1993-1994 levels. The rise in immigration enforcement can therefore explain three-quarters of the actual 4.4 percentage point decline during this time. A large fraction of the decline in immigrant Medicaid participation around 1996, which has previously been attributed to welfare reform, is due to the contemporaneous rise in immigration enforcement.

VII. Conclusion.

The results presented here cast new light on the chilling of immigrant Medicaid participation around the time of welfare reform. Previous literature documents such chilling and hypothesizes that the decline in participation stems from fear and confusion stemming from changes in welfare policy. This research suggests a new potential culprit - Federal immigration enforcement – which contributed to immigrant reluctance to

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²⁸ These results are sensitive to the choice of years. Using 1999 as the relevant survey year for the end of the window leads to the conclusion that enforcement explains 40 percent of the decline.

participate in Medicaid. The results imply that a majority of the decline in immigrant Medicaid participation around the time of welfare reform can in fact be attributed to increased enforcement of immigration law.

Future versions of this paper will also incorporate media coverage of enforcement actions in Spanish- and English-language newspapers.²⁹ Media coverage will serve as a proxy for the degree of awareness about the level of enforcement in a specific area, offering a deeper understanding of how enforcement affects policy actions.

The results of this paper point to a potentially unintended consequence of immigration policy, and highlight the fact that seemingly unrelated policies can have important consequences for program take-up.

²⁹ Data collection using the Proquest Ethnic NewsWatch and Lexis Nexis databases is nearly complete.

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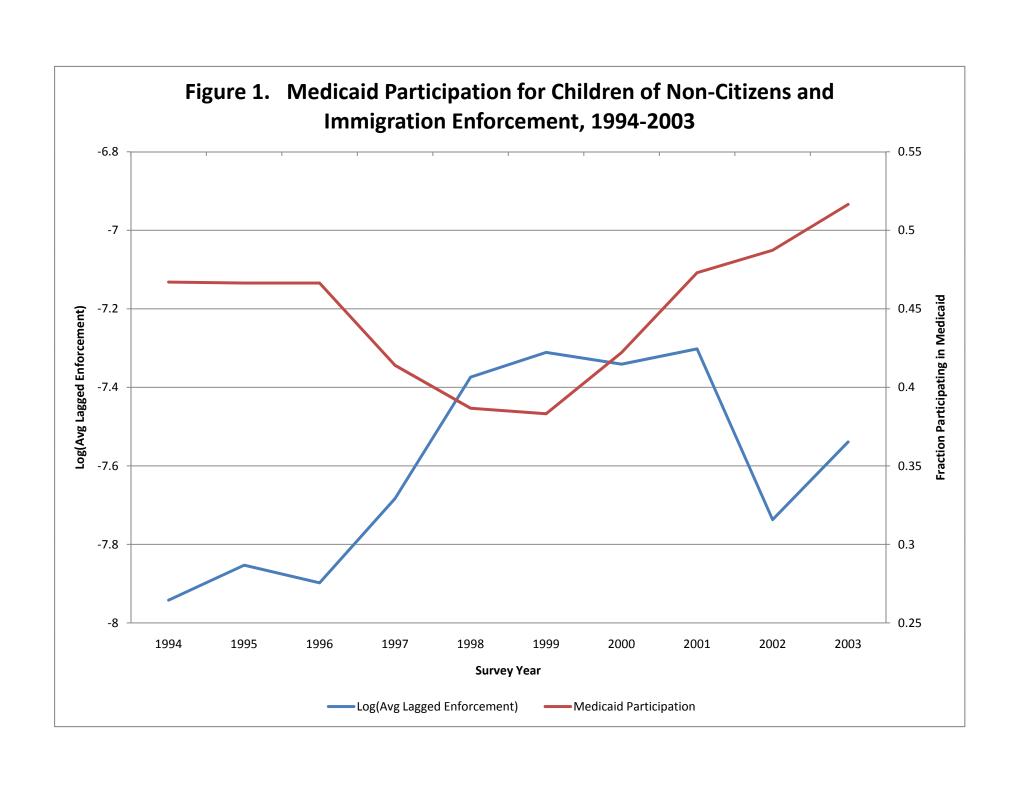
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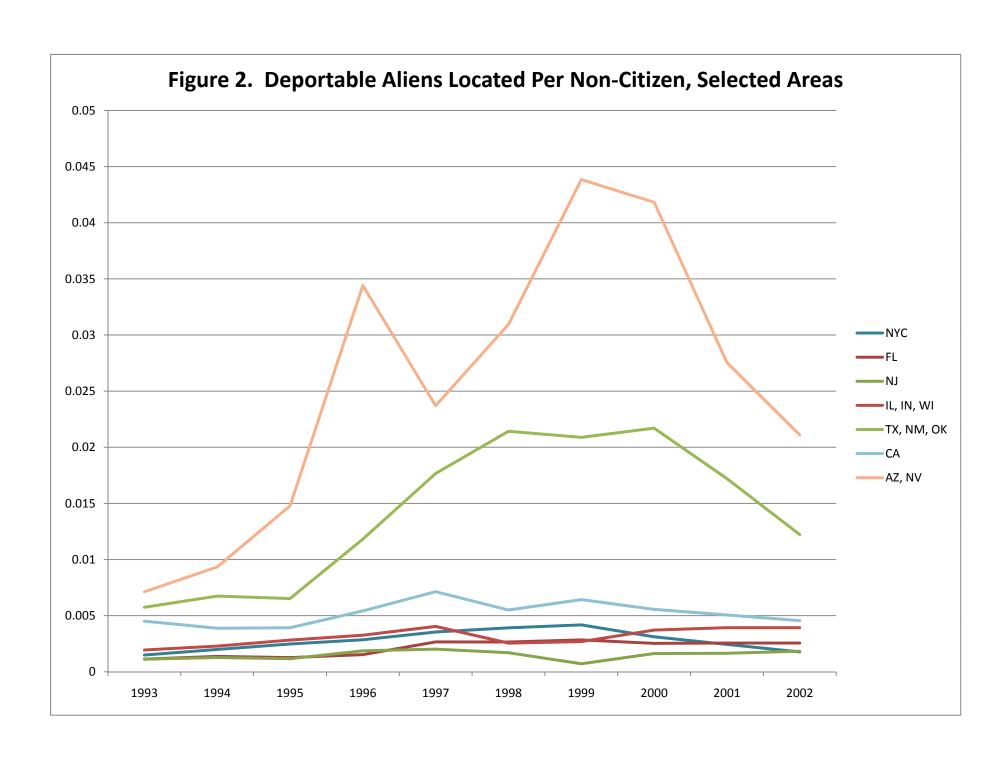


Table 1. Means of Key Variables

Low SES Sample

Full Sample

		Mom Foreign Born			Mom Foreign Born	
	Mom Non-Citizen	Citizen	Mom Native Born	Mom Non-Citizen	Citizen	Mom Native Born
	(N=26,942)	(N=6,978)	(N=112,286)	(N=42,012)	(N=19,371)	(N=331,558)
Medicaid	0.45	0.39	0.47	0.32	0.17	0.20
Medicaid Eligible (Definition 1)	0.80	0.75	0.73	0.54	0.31	0.29
Medicaid Eligibile (Definition 2)	0.69	0.67	0.64	0.47	0.28	0.26
Any Health Insurance	0.68	0.74	0.84	0.74	0.85	0.90
Any Food Stamps	0.27	0.23	0.35	0.17	0.09	0.13
Any Public Assistance/Welfare	0.17	0.14	0.22	0.11	0.05	0.08
Child is Citizen	0.81	0.96	1.00	0.82	0.97	1.00
Mom is High School Grad Exactly	0.22	0.33	0.45	0.22	0.27	0.34
Mom is Some College Exactly	0.09	0.22	0.29	0.13	0.24	0.31
Mom is College Grad or More	0.00	0.00	0.00	0.13	0.27	0.23
Family Under 200% FPL	1.00	1.00	1.00	0.65	0.37	0.36
Mom worked Last Year	0.42	0.57	0.62	0.52	0.73	0.75
Mom Married	0.76	0.70	0.49	0.82	0.84	0.75
Mom Spouse Citizen (if married)	0.22	0.70	0.96	0.31	0.81	0.98
Lives in Generous State (Borjas Definition)	0.89	0.88	0.68	0.89	0.89	0.71
Lives in Generous State (KK Definition)	0.57	0.49	0.41	0.56	0.53	0.45
Lives in Generous State (ZT Definition)	0.72	0.70	0.44	0.72	0.75	0.48
Anti-Immigrant Sentiment in State	0.53	0.52	0.53	0.53	0.52	0.53
Child Age	7.47	9.08	7.91	7.51	9.23	8.45
Enforcement Level in Cluster*1000	7.79	7.80	8.44	7.53	6.92	8.39
Enforcement Level for Group*1000	1.99	0.88	n/a	1.61	0.49	n/a
Enforcement Level in Cluster-Group*1000	2.15	1.11	n/a	1.73	0.60	n/a

Notes: All samples exclude non-citizen children whose mothers arrived within the past five years. The Low-SES sample includes children of mothers lacking a college degree and under 200 percent of the poverty line. Medicaid eligibility definition 1 imputes the AFDC/TANF eligibility pathway; Medicaid eligibility definition 2 does not. Measures of state generosity and anti-immigrant sentiment described in text. Enforcement level is the average number of deportable aliens located in the reference year and previous year per non-citizen in the cluster, group, or cluster-group.

Table 2. Preliminary Analysis

Dep.Var: Medicaid Participation				
Sample	Mother Non-Citizen	Mother Non-Citizen	Mother Citizen	Mother Citizen
Cluster f.e.	yes	yes	yes	yes
Year f.e.	yes	yes	yes	yes
Cluster-specific time trends	yes	yes	yes	yes
Panel A. Low SES Sample				
Log(Enforcement)	-0.087+ (0.043)		0.019 (0.022)	
Log(Lead of Enforcement)	(0.005 (0.032)	(/	-0.002 (0.023)
Number of Observations	26942	23528	119264	102790
R-squared	0.042	0.041	0.016	0.016
Panel B. Full Sample				
Log(Enforcement)	-0.049+ (0.027)		0.014 (0.014)	
Log(Lead of Enforcement)	(====)	0.013 (0.018)	(,	0.002 (0.013)
Number of Observations	42012	36103	350929	296775
R-squared	0.033	0.033	0.010	0.010

Table 3. Effect of Immigration Enforcement on Medicaid Participation

Dep.Var: Medicaid Participation						
					Kid Citizen and	
				Mom Arrived > 5	Mom Arrived> 5	Mom Foreign
Sample	All	All	Kid Citizen	Years	years	Born
Mom Non-Cit*State f.e.	yes					
Mom Non-Cit*Year f.e.	yes	yes	yes	yes	yes	yes
Log(Enforcement)	yes	yes	yes	yes	yes	yes
State*Group*Mom Non-Cit f.e.		yes	yes	yes	yes	yes
Demographic Controls		yes	yes	yes	yes	yes
Panel A. Low SES Sample						
Mom Non-Cit*Log(Enforcement)	-0.106**	-0.099*	-0.102*	-0.098**	-0.100*	-0.110*
	(0.031)	(0.036)	(0.044)	(0.035)	(0.043)	(0.050)
Number of Observations	146,206	146,206	140,587	143,599	137,980	33,920
R-squared	0.025	0.101	0.098	0.099	0.096	0.192
Panel B. Full Sample						
Mom Non-Cit*Log(Enforcement)	-0.049*	-0.053*	-0.052+	-0.052*	-0.052+	-0.083*
	(0.023)	(0.021)	(0.025)	(0.022)	(0.026)	(0.030)
Number of Observations	392,941	392,939	384,288	388,856	380,205	61,383
R-squared	0.023	0.279	0.283	0.279	0.282	0.279

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. New York City and the remainder of New York are treated as distinct states because they lie in different INS clusters. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster. Demographic controls include dummies for mother's educational attainment, age*year fixed effects, and (in Panel B) an indicator for being below 200 percent of the poverty line.

Table 4. Effect of Immigration Enforcement on Medicaid Eligibility

Mom Non-Cit*Year f.e. yes yes yes Log(Enforcement) yes yes yes State*Group*Mom Non-Cit f.e. yes yes yes Demographic Controls yes yes yes Panel A. Low SES Sample Mom Non-Cit*Log(Enforcement) -0.099* 0.013 0.029 Number of Observations 146,206 146,206 146,206 R-squared 0.101 0.238 0.427 Panel B. Full Sample -0.053* 0.011 0.018 Number of Observations 392,939 392,939 392,939 Number of Observations 392,939 392,939 392,939 R-squared 0.279 0.585 0.543	Dependent Variable	Medicaid Participation	Eligibility Def 1	Eligibility Def2
State*Group*Mom Non-Cit f.e. yes pes yes yes yes yes yes yes	Mom Non-Cit*Year f.e.	yes	yes	yes
Panel A. Low SES Sample Yes yes yes Mom Non-Cit*Log(Enforcement) -0.099* (0.036) 0.013 (0.031) 0.029 (0.033) Number of Observations R-squared 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206	Log(Enforcement)	yes	yes	yes
Panel A. Low SES Sample Yes yes yes Mom Non-Cit*Log(Enforcement) -0.099* (0.036) 0.013 (0.031) 0.029 (0.033) Number of Observations R-squared 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206 (0.031) 146,206	State*Group*Mom Non-Cit f.e.	yes	yes	yes
Mom Non-Cit*Log(Enforcement) -0.099* (0.036) 0.013 (0.031) 0.029 (0.033) Number of Observations 146,206 146,206 146,206 R-squared 0.101 0.238 0.427 Panel B. Full Sample -0.053* (0.021) 0.011 (0.027) 0.018 (0.026) Number of Observations 392,939 392,939 392,939		yes	yes	yes
Number of Observations 146,206 146,206 146,206 R-squared 0.101 0.238 0.427 Panel B. Full Sample Mom Non-Cit*Log(Enforcement) -0.053* 0.011 0.018 (0.021) (0.027) (0.026) Number of Observations 392,939 392,939 392,939	Panel A. Low SES Sample			
Number of Observations 146,206 146,206 146,206 R-squared 0.101 0.238 0.427 Panel B. Full Sample Mom Non-Cit*Log(Enforcement) -0.053* 0.011 0.018 (0.021) (0.027) (0.026) Number of Observations 392,939 392,939 392,939	Mom Non-Cit*Log(Enforcement)	-0.099*	0.013	0.029
R-squared 0.101 0.238 0.427 Panel B. Full Sample -0.053* 0.011 0.018 (0.021) (0.027) (0.026) Number of Observations 392,939 392,939 392,939	wom non ele 205/2 more ement,	(0.036)	(0.031)	(0.033)
R-squared 0.101 0.238 0.427 Panel B. Full Sample -0.053* 0.011 0.018 (0.021) (0.027) (0.026) Number of Observations 392,939 392,939 392,939	Number of Observations	146 206	146 206	146 206
Mom Non-Cit*Log(Enforcement) -0.053* (0.021) 0.011 (0.027) 0.018 (0.026) Number of Observations 392,939 392,939 392,939				
Mom Non-Cit*Log(Enforcement) -0.053* (0.021) 0.011 (0.027) 0.018 (0.026) Number of Observations 392,939 392,939 392,939				
(0.021) (0.027) (0.026) Number of Observations 392,939 392,939 392,939	Panel B. Full Sample			
Number of Observations 392,939 392,939 392,939	Mom Non-Cit*Log(Enforcement)	-0.053*	0.011	0.018
		(0.021)	(0.027)	(0.026)
	Number of Observations	202 020	392 939	392 939
	R-squared	0.279	0.585	0.543

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. New York City and the remainder of New York are treated as distinct states because they lie in different INS clusters. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster. Demographic controls include dummies for mother's educational attainment, age*year fixed effects, and (in Panel B) an indicator for being below 200 percent of the poverty line. Eligibility Definition 1 imputes eligibility incorporating the AFDC/TANF eligibility pathway.

Table 5. Effect of Immigration Enforcement on Take-Up (Participation Conditional on Imputed Eligibility)

Dependent Variable	/ariable Medicaid		Medicaid (if Eligibile by Definition 2)
Mom Non-Cit*Year f.e. Log(Enforcement)	yes	yes	yes
	yes	yes	yes
State*Group*Mom Non-Cit f.e. Demographic Controls	yes	yes	yes
	yes	yes	yes
Panel A. Low SES Sample			
Mom Non-Cit*Log(Enforcement)	-0.099*	-0.117*	-0.099*
	(0.036)	(0.048)	(0.041)
Number of Observations	146,206	109,433	96,444
R-squared	0.101	0.098	0.098
Panel B. Full Sample			
Mom Non-Cit*Log(Enforcement)	-0.053*	-0.102*	-0.080+
	(0.021)	(0.047)	(0.040)
Number of Observations	392,939	126,839	112,367
R-squared	0.279	0.133	0.134

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. New York City and the remainder of New York are treated as distinct states because they lie in different INS clusters. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster. Demographic controls include dummies for mother's educational attainment, age*year fixed effects, and (in Panel B) an indicator for being below 200 percent of the poverty line.

Table 6. Measurement of Enforcement, Foreign-Born Low SES Sample

Dep.Var: Medicaid Participation	İ	II	III	IV	V	VI	VII	VIII	IX	Х	XI	XII
Sample	All	All	All	All	Low	Low	Low	Low	High	High	High	High
					Exposure							
Mom Non-Cit*Log(Enforcement of All Non-Citizens in Cluster)	-0.110*			-0.126+	-0.085			-0.043	-0.245**			-0.316**
	(0.050)			(0.066)	(0.068)			(0.103)	(0.078)			(0.067)
Mom Non-Cit*Log(Enforcement of Group Nationally)		0.066		0.087		0.006		-0.037		0.062		0.079+
		(0.106)		(0.086)		(0.136)		(0.178)		(0.077)		(0.043)
Mom Non-Cit*Log(Enforcement of Group Within Cluster)			-0.043	-0.001			0.010	0.065			-0.118**	-0.031
			(0.043)	(0.047)			(0.050)	(0.096)			(0.040)	(0.047)
Number of Observations	33920	33920	33809	33809	12089	12089	12054	12054	16475	16475	16457	16457
R-squared	0.192	0.192	0.190	0.190	0.263	0.262	0.259	0.261	0.172	0.171	0.170	0.172

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. All regressions include full set of fixed effects and demographic controls as in the previous tables. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster. High exposure indicates that the child resides in an area with greater than or equal to the fraction of non-citizens of a typical child from her (mother's) country of origin group.

Table 7. Instrumenting for Citizenship (Low SES Sample)

Dep.Var: Medicaid Participation	OLS	IV
Mom Non-Cit*Year f.e.	yes	Instrumented
Log(Enforcement)	yes	yes
State*Group f.e.	yes	yes
Demographic Controls	yes	yes
	0.04544	0.405**
Mom Non-Cit*Log(Enforcement)	-0.045**	-0.106**
	(0.009)	(0.040)
Number of Observations	146206	146206
R-squared	0.095	

Notes: OLS model differs from baseline model in that state-group fixed effects are included rather than state-group-citizen fixed effects. IV model instruments for citizen*year fixed effects and citizen*enforcement using group*year and group*enforcement. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. Sample excludes non-citizen children whose mothers arrived within five years and is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. All regressions include full set of fixed effects and demographic controls as in the previous table. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster.

Table 8. Effect of Immigration Enforcement on Insurance Status and Program Participation

Dependent Variable	Medicaid	Public Health Insuance	Private Health Insurance	Any Health Insurance	Public Assistance/ Welfare	Food Stamps
Mom Non-Cit*Year f.e.	yes	yes	yes	yes	yes	yes
Log(Enforcement)	yes	yes	yes	yes	yes	yes
State*Group*Mom Non-Cit f.e. Demographic Controls	yes yes	yes yes	yes yes	yes yes	yes yes	yes yes
Panel A. Low SES Sample						
Mom Non-Cit*Log(Enforcement)	-0.099* (0.036)	-0.098** (0.034)	0.017 (0.026)	-0.060* (0.022)	-0.022 (0.019)	-0.084+ (0.041)
Number of Observations R-squared	146,206 0.101	146,206 0.096	146,206 0.105	146,206 0.081	146,206 0.132	146,206 0.091
Panel B. Full Sample						
Mom Non-Cit*Log(Enforcement)	- 0.053* (0.021)	- 0.053* (0.021)	- 0.010 (0.019)	-0.044* (0.017)	0.002 (0.009)	- 0.055 (0.033)
Number of Observations R-squared	392,939 0.279	392,939 0.252	392,939 0.338	392,939 0.095	392,939 0.183	392,939 0.134

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. New York City and the remainder of New York are treated as distinct states because they lie in different INS clusters. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster. Demographic controls include dummies for mother's educational attainment, age*year fixed effects, and (in Panel B) an indicator for being below 200 percent of the poverty line.

Table 9. State-Level Determinants of Participation, Low SES sample

Dep.Var: Medicaid Participation

Measure of State Climate	Anti-Immigrant Sentiment	Generous: Borjas definition	Generous: Kaushal and Kaestner definition	Generous: Zimmerman and Tumlin definition	Anti- Immigrant Sentiment	Borjas definition	Kaushal and Kaestner definition	Zimmerman and Tumlin definition
Mom Non-Cit*Log(Enforcement)					-0.104**	-0.118**	-0.120**	-0.068*
					(0.036)	(0.034)	(0.030)	(0.027)
Mom Non-Cit*State Climate	0.007	-0.050	-0.007	0.072	0.024	-0.096	-0.045	0.051
	(0.161)	(0.045)	(0.047)	(0.043)	(0.161)	(0.056)	(0.038)	(0.041)
Number of Observations	143,244	146,206	146,206	146,206	143,244	146,206	146,206	146,206
R-squared	0.100	0.100	0.100	0.100	0.101	0.101	0.101	0.101

^{*}describe geberosity here**Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. All regressions include full set of fixed effects and demographic controls as in the previous table. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster.

Appendix Table 1. Does Enforcement Predict Other Characteristics?

Dependent Variable	Poverty Under 200% FPL	Poverty Under 100% FPL	Mom Married	Mom Spouse Citizen (if Married)	Mom College Grad	Mom Some College	Mom High School Grad	Mom Worked Last Year	Age of Child	Mom Arrived Within Five Years
Panel A. Low SES Sample										
Mom Non-Cit*Log(Enforcement)	n/a	-0.004	0.031	-0.006	n/a	-0.007	0.012	0.005	-0.174	-0.003
		(0.029)	(0.022)	(0.018)		(0.017)	(0.023)	(0.022)	(0.238)	(0.009)
Number of Observations R-squared		146,206 0.045	146,206 0.093	80,038 0.634		146,206 0.081	146,206 0.078	146,206 0.081	146,206 0.028	146,206 0.200
Panel B. Full Sample										
Mom Non-Cit*Log(Enforcement)	0.004 (0.023)	-0.014 (0.024)	-0.025 (0.018)	0.019 (0.015)	-0.015 (0.015)	0.005 (0.015)	0.016 (0.015)	0.015 (0.027)	-0.101 (0.176)	0.004 (0.009)
Number of Observations R-squared	392,941 0.095	392,941 0.066	392,941 0.037	297,360 0.563	392,939 0.068	392,939 0.046	392,939 0.037	392,939 0.063	392,941 0.021	392,941 0.192

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. All regressions include full set of fixed effects and demographic controls as in the previous table. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster.

Appendix Table 2. Robustness Checks, Low SES Sample

Dependent Variable: Medicaid

Change from Preferred Specification	Preferred Specification	Linear Enforcement Measure	Restrict to Moms Arriving Before 1992	Add State-Specific Linear Time Trends
Sample	All	All	All Arriving Before 1992	All
Mom Non-Cit*Log(Enforcement)	-0.099*		-0.096*	-0.105**
Mom Non-Cit*Enforcement	(0.036)	-5.859* (2.667)	(0.035)	(0.036)
Number of Observations R-squared	146206 0.101	146206 0.101	138897 0.098	146206 0.112

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. All regressions include full set of fixed effects and demographic controls as in the previous table. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster.

Appendix Table 3. Alternative Definitions of Citizenship, Low SES Sample

Dependent Variable: Medicaid

Definition of Citizenship	Mom Non- Citizen	Mom's Spouse Non-Citizen	Mom and Spouse Non- Citizen	Mom and Spouse Mixed Citizenship	Any Parent Non-Citizen	Any Parent Non-Citizen	Household Head Non- Citizen	Mom's Spouse Non-Citizen	Mom's Spouse Non-Citizen
Sample	All	Married	Married	Married	Married	All	All	Married and Non-Citizen Mom	Married and Citizen Mom
Definition of Non-Cit*Log(Enforcement)	-0.099*	-0.123**	-0.120**	-0.085+	-0.112**	-0.105**	-0.113**	-0.055	-0.156**
	(0.036)	(0.032)	(0.037)	(0.044)	(0.038)	(0.031)	(0.027)	(0.034)	(0.051)
Number of Observations R-squared	146206	80038	80038	80038	80038	146206	146206	19857	60181
	0.101	0.109	0.106	0.108	0.108	0.102	0.102	0.201	0.092

Notes: Linear probability model. Standard errors in parantheses are clustered by INS cluster. **, *, and + represent statistical significance at the 1, 5, and 10 percent levels. All samples exclude non-citizen children whose mothers arrived within five years. The Low SES sample is restricted to children living below 200 percent of the poverty line and whose mothers have less than a college degree. All regressions include full set of fixed effects and demographic controls as in the previous tables, except that citizenship categories for fixed effects are defined as indicated. Enforcement is measured as the average of the number of deportable aliens located per non-citizen in the reference year and the year prior to the reference year in the INS cluster.