The EITC and the Extensive Margin: A Reappraisal*

Henrik Kleven Princeton University and NBER

May 2024

Abstract

A strong consensus posits that the Earned Income Tax Credit (EITC) has had sizable effects on extensive margin labor supply, especially for single mothers. This paper reappraises the difference-in-differences and event study approaches that underpin much of this consensus. The paper investigates every EITC reform at the state and federal level since the inception of the policy. All reforms are analyzed in an event study framework and a comprehensive analysis of model uncertainty is presented. Apart from the federal 1993 reform, EITC expansions are not associated with any clear and robust effects on employment. Treatment impact estimates from about 500 event studies are symmetrically distributed around zero. Specifications with large elasticities are outliers in the distribution. The 1993 reform, on the other hand, is associated with large employment increases for single mothers. Based on a number of different analyses, the paper shows that these increases align more closely with confounding effects from welfare reform and the macroeconomy than with the EITC. Overall, difference-in-differences and event study analyses of EITC reforms are fragile to specification choices and do not support robustly large effects.

^{*}I thank Richard Blundell, Raj Chetty, Monica Costa Dias, Arindrajit Dube, Amy Finkelstein, John Friedman, Peter Ganong, Alex Gelber, Jeffrey Grogger, Nathan Hendren, Hilary Hoynes, Theodore Joyce, Larry Katz, Pat Kline, Wojciech Kopczuk, Ilyana Kuziemko, Claus Kreiner, Camille Landais, Thomas Lemieux, Adam Looney, Day Manoli, Alex Mas, Maxim Massenkoff, Larry Mead, Costas Meghir, Bruce Meyer, Pascal Noel, Matthew Notowidigdo, Jesse Rothstein, Dominik Sachs, Emmanuel Saez, Stefanie Stantcheva, Michael Wiseman, Owen Zidar, and Jim Ziliak for comments and discussions. I am grateful to Eva Demsky, Ragini Jain, Madhavi Jha, Ale Marchetti-Bowick, Aarti Malik, and Shreya Tandon for outstanding research assistance, and to Stephanie Kestelman and Dana Scott for their assistance at an early stage of the project. Contact information for the author: Henrik Kleven, Department of Economics and the School of Public and International Affairs, Princeton University, web: www.henrikkleven.com, e-mail: kleven@princeton.edu.

1 Introduction

Anti-poverty policy in the United States has changed fundamentally over the last fifty years. One of the most profound changes has been the expansion of the Earned Income Tax Credit (EITC) along with the downsizing of traditional cash welfare, thus shifting low-income support from the poor to the working poor. The EITC program now dwarfs cash welfare by a factor of almost 30 in terms of the number of recipients.¹ A key motivation for this policy shift has been to encourage labor supply at the extensive margin. Traditional welfare has long been blamed for keeping families out of the workforce (e.g., Murray 1984), while the EITC is supposed to draw them in.

A large literature on the EITC features a striking degree of consensus. Most authors argue that the program has had sizable effects on extensive margin labor supply, especially for single mothers. Starting with the important contributions by Eissa and Liebman (1996) and Meyer and Rosenbaum (2001), the literature has focused mostly on the federal EITC reforms in the 1980s and 1990s and relied on difference-in-differences approaches using variation by the presence and number of children. The most striking evidence in support of extensive margin responses is the large increase in the employment rate of single mothers, especially single mothers with two or more children, following the 1993 EITC expansion for these family types (e.g., Meyer 2010; Hoynes and Patel 2018).

In this paper, I study the robustness of such difference-in-differences approaches to estimating the effects of the EITC on extensive margin labor supply. The analysis has four key elements. First, I take a long-run perspective and consider every EITC reform at the federal and state level since the inception of the policy in 1975. Second, I analyze all reforms in an event study framework, looking at the dynamics of labor supply changes before and after legislated reforms. Third, I provide a comprehensive study of model uncertainty, comparing prior EITC estimates to the distribution of estimates under different empirical models. Fourth, I investigate the role of confounding factors, especially for the 1993 reform which underpins much of the consensus. The previous literature also considered the role of confounders in the 1990s, but I provide new and transparent analyses of the issue.

Like most of the existing literature, I use Current Population Survey (CPS) data and focus on

¹See Figure A.1 in the online appendix.

the labor supply of single mothers.² For these women, the extensive margin incentives created by the EITC are unambiguously positive. I start from a long-run historical perspective, documenting the evolution in labor force participation of single women with and without children over the last 50 years. The observed patterns are striking. The participation rates for these two groups have evolved in parallel over this entire period, except for the mid-late 1990s. The participation rate for those with children was about 14 percentage points lower than for those without children in the late 1960s. This gap was about the same 25 years later, in the early 1990s. Then single mothers closed the entire gap in about five years, after which the two groups reverted to parallel trends. The dramatic increase for single mothers followed the 1993 EITC expansion, but the long-run series highlight that this period was an anomaly. If the EITC was important for the extensive margin in the 1990s, then why did it not narrow the gap in other time periods? There were federal EITC reforms in 1975, 1986, 1990 and 2009 along with numerous state EITC reforms between 1984-2018, which significantly increased the incentive to enter the labor market for single mothers.

Following the descriptive long-run evidence, I present event studies of EITC reforms at the federal and state levels. For federal reforms, the analyses are based on difference-in-differences specifications comparing single women with and without children, or single women with difference specifications comparing single mothers in states are based on either difference-in-differences specifications comparing single mothers in states with and without EITC reform or triple-differences specifications comparing single women with and without children in states with and without EITC reform. Apart from the 1993 federal reform, the event studies are essentially flat around the various EITC expansions. This holds in the sample of all single mothers and in samples of single mothers more strongly treated by the EITC. For the 1993 federal reform, on the other hand, the estimated effects are consistently positive and large.

Empirical designs that define treatment status based on fertility rest on strong assumptions, because the impact of children on labor market outcomes is very large (see Kleven, Landais, and Søgaard 2019; Kleven 2023). This creates significant imbalance in the levels of pre-reform outcomes for treatment and control groups. Investigating pre-trends is useful for assessing the validity of the difference-in-differences approach and these generally look good, but concerns remain that contemporaneous shocks affect treatment and control groups differently. Hence, the central challenge for estimating causal impacts lies in specifying an empirical model that avoids bias from

²While the existing literature has relied primarily on the March supplement of the CPS, I use linked March and monthly files. The larger dataset is very useful for precision in some of the more demanding event study specifications.

confounding shocks. This paper takes an agnostic approach to the issue of model uncertainty, estimating EITC impacts across a wide range of specifications. These specifications consider different reform episodes, different samples, different comparison groups, different extensive margin measures, and different control variables. Allowing for all possible permutations of these specification choices yield hundreds of treatment impact estimates. Leaving aside the 1993 federal reform, the estimates are symmetrically distributed around zero. I show that prior estimates are strong outliers in the distribution of possible estimates. For the 1993 reform, on the other hand, the distribution of estimates is shifted to the right and has a mean elasticity of 0.63.

Any compelling narrative regarding the EITC and the extensive margin must reconcile the starkly different patterns around 1993 and elsewhere. One view is that the 1993 reform was different simply by virtue of being larger than other reforms. In a world with optimization frictions, elasticities may depend on the size of the tax variation used to estimate them (Chetty, Friedman, Olsen, and Pistaferri 2011; Chetty 2012; Kleven 2016). The problem with this interpretation is that, while the 1993 reform was larger than other reforms for families with two or more children, it was not larger for families with one child. For example, the 1975 EITC introduction was larger than the 1993 expansion for one-child families, and yet we observe clear employment increases for this subgroup only around the 1993 reform. Another view is that the anomalous patterns of the 1990s were driven, not by the EITC, but by confounding factors. The importance of such factors has been discussed in the literature (e.g., Ellwood 2000; Meyer and Rosenbaum 2001; Blank 2002; Grogger 2003; Fang and Keane 2004), with authors arguing that the EITC was in fact a major component of what happened in the 1990s even if it wasn't solely responsible.

What were the main confounders in the 1990s? First, there was welfare reform. This includes a large number of state welfare reforms (so-called welfare waivers) between 1992-96 and the federal welfare reform act of 1996. These policy changes introduced constraints on welfare receipt such as time limits and work requirements, pushing single mothers off cash welfare and producing extensive margin incentives correlated with the EITC. Second, there was a booming macroeconomy. Business cycles may have heterogeneous effects on single women with and without children, especially at a time where welfare reform puts strong pressure on single mothers to find employment. Third, changes to social norms may have played a role as well. I provide evidence that attitudes towards welfare receipt and work changed dramatically in the 1990s — the passage of welfare reform arguably grew out of those attitude changes — and such social norms could have had an independent effect on behavior.

Based on a detailed investigation of confounders in the 1990s, I argue that the empirical patterns align more closely with confounders than with EITC reform. Let me outline some of the key findings. First, simulation exercises reveal that the fraction of the employment increase for single mothers attributable to the 1993 EITC expansion is small even under large extensive margin elasticities. In other words, within any reasonable range of tax elasticities, the patterns around the 1993 reform must be driven mostly by confounders and not by the EITC. Second, the extensive margin effects for single mothers following the 1993 reform are strongly increasing in the number of children.³ Such fanning-out by family size is exactly consistent with welfare reform, because pre-reform welfare participation and therefore welfare treatment was strongly increasing in family size. EITC reform, on the other hand, is not consistent with this pattern. Third, using a rich set of demographic variables to predict welfare treatment intensity, I show that the extensive margin effects in the 1990s are driven precisely by those most treated by welfare reform.⁴ EITC-eligible women who were only weakly affected by welfare reform did not increase labor supply at the extensive margin. Finally, I show that the employment effects between 1994 and 1996 — after the 1993 EITC reform, but before federal welfare reform — can be explained by state welfare waivers and the business cycle. Controlling for business cycle effects, non-waiver states did not see any significant increase in the employment of single women with children relative to those without prior to federal welfare reform.

To summarize, this paper shows that difference-in-differences and event study approaches to analyzing EITC reform are fragile. In contrast to the current consensus, large extensive margin responses are not a robust feature of the data. Based on a multitude of EITC reforms and empirical analyses, I find that the estimates of treatment impacts are clustered around zero, except for the strongly confounded 1993 reform. Importantly, these results capture only the observable short-run responses to the EITC. It is possible that long-run responses are larger than short-run responses due to optimization frictions or labor supply dynamics. In general, difference-in-differences and event study designs are not well-suited for estimating long-run effects in the presence of such mechanisms (Kleven, Kreiner, Larsen, and Søgaard 2023). This limitation applies equally to my paper and the previous literature.⁵

³After the 1993 reform, the employment rate of single mothers (relative to single childless women) increased by about 10pp for those with one child, 15pp for those with two children, 20pp for those with three children, and almost 30pp for those with four or more children.

⁴To predict welfare treatment intensity, I use that pre-reform welfare participation rates (and the ensuing drop in welfare caseloads) are strongly related to demographics such as the number of children, the age of the youngest child, the age of the mother, race, and state.

⁵It is worth noting that the long-run descriptive evidence discussed above is not obviously suggestive of slow-

Can these findings be reconciled with the broader literature on labor supply responses to economic incentives? Let us consider quasi-experimental studies and randomized evaluations in turn. While my conclusions are at odds with prior quasi-experimental studies of EITC reform, they are not necessarily at odds with quasi-experimental studies of tax reform more broadly. After decades of research, we have surprisingly little evidence showing clear and sizeable effects of taxes on real labor supply.⁶ It is telling that the meta study of extensive margin elasticities by Chetty, Guren, Manoli, and Weber (2013) cites only nine studies of (Hicksian) elasticities, most of which are relatively old and not based on modern quasi-experimental designs. Studies that do find sizeable behavioral responses to tax reform mostly come from the literature on taxable income responses at the intensive margin, driven by avoidance and evasion behavior (Saez, Slemrod, and Giertz 2012). Therefore, prior EITC studies are arguably stronger outliers in the literature on behavioral responses to taxes than the study presented here.

A number of randomized evaluations do find extensive margin effects of work incentives. Most of these evaluations consider the impact of welfare reform treatments such as time limits, work requirements, and various financial incentives, often bundled together. These are precisely the aspects that featured prominently in federal welfare reform and state welfare waivers, and in fact most of the trials were demonstration projects for waiver-based reform (Grogger and Karoly 2005; Kline and Tartari 2016).⁷ Finding extensive margin effects in such welfare trials is consistent with the arguments regarding welfare reform presented here.

As for in-work benefits like the EITC, we have very few randomized evaluations. Card and Hyslop (2005) is an important exception. They evaluate an earnings subsidy to welfare recipients in Canada, finding effects in the short run but not in the long run. However, this policy was structured differently from the EITC and involved strong intertemporal substitution incentives.⁸

moving responses resulting from optimization frictions. Standard theories of frictions would predict gradual convergence between treatment and control groups as more women re-optimize labor supply over time. Instead the two groups have moved in parallel over the last 50 years, apart from the brief and dramatic changes in the mid-late 1990s. Reconciling such patterns with the idea of sizable long-run elasticities requires a particular model of frictions (presumably some form of tipping-point model).

⁶Conversely, we *do* have clear examples of precisely estimated zeros. This includes evidence from a two-year income tax holiday in Switzerland (Martinez, Saez, and Siegenthaler 2021) and from a cohort-based payroll tax reform in Greece (Saez, Matsaganis, and Tsakloglou 2012). Both of these papers find zero effects of taxes on extensive margin labor supply. The payroll tax paper is particularly informative because the cohort-based feature of the reform allows for estimating long-run responses.

⁷Grogger and Karoly (2005) provide a comprehensive review of these waiver demonstration trials. Kline and Tartari (2016) provide a recent evaluation of Connecticut's Jobs First Program, a waiver demonstration that included time limits, work requirements, and financial incentives. They find substantial extensive margin labor supply effects.

⁸As a result, Chetty, Guren, Manoli, and Weber (2013) use this experiment to back out the Frisch elasticity, as opposed to the Hicksian (steady state) elasticity that is the focus of this paper.

Recently, two studies provide evidence from the Paycheck Plus demonstration in New York City (Miller, Katz, Azurdia, Isen, Schultz, and Aloisi 2018) and Atlanta (Yang, Bernardi, Metz, Miller, Katz, and Isen 2022). The Paycheck Plus program is similar to the EITC, but targeted to workers without dependent children. The New York City trial had modest employment effects, while the Atlanta trial had no significant employment effects. The findings from these trials — the closest we have to actual EITC experiments — are similar to the findings in this paper: the employment effects are at most modest.⁹ The fact that the actual EITC is likely associated with less information than Paycheck Plus reinforces this conclusion. A considerable amount of evidence shows that potential EITC recipients are often unaware of the program or have a limited understanding of its complicated schedule, eligibility and claiming rules (e.g., Bhargava and Manoli 2015). All else equal, this suggests that Paycheck Plus provides an upper bound on real-world EITC effects.¹⁰

How should we interpret the starkly different effects of EITC and welfare reform estimated in this paper? Three interpretations are worth highlighting. The first is that income effects may be important. Unlike tax credits, the removal of welfare benefits creates positive income effects on extensive margin labor supply: single mothers had to find work to maintain their standard of living. The second interpretation is that EITC and welfare reform are associated with different informational and psychological frictions. The welfare debate was much more salient than the EITC debate in the 1990s. Moreover, welfare reform used fundamentally different policy instruments — ordeals and enforcement — than EITC reform. The implications of these instruments were transparent and largely mechanical, unlike the implications of a complicated, nonlinear tax refund. The third interpretation is that welfare reform was associated with a change in social norms regarding welfare dependence, thus creating a social multiplier effect on labor supply. All of these mechanisms may have been at play and jointly explain the larger effects of welfare reform.¹¹

The rest of the paper is organized as follows. Section 2 reviews the related literature. Section 3 describes the policy context and data. Section 4 presents long-run historical evidence. Sections 5 and 6 provide quasi-experimental evidence on the impact of EITC reforms at the federal and state

⁹As I show in the paper, the elasticities implied by the quasi-experimental EITC literature are an order of magnitude larger than the elasticities implied by these randomized trials.

¹⁰Of course, the comparison may be confounded by other external validity concerns, including the fact that Paycheck Plus targets a different population.

¹¹It is worth noting that this paper (as well as the existing literature) tries to separately identify the effects of EITC and welfare reform by treating the two policies as additively separable. It is conceivable that the treatment effects interact such that the *combination* of EITC expansion and welfare contraction was critical for the large employment effects in the 1990s. This is still consistent with the main conclusion of the paper: the EITC program on its own has not had robustly large employment effects; the implementation of welfare reform (and the macroeconomy) was necessary for the patterns observed in the 1990s.

levels. Section 7 dissects the empirical patterns of the 1990s, trying to separate EITC effects from the effects of confounding factors. Section 8 concludes.

2 Literature Review

An enormous body of work has examined the labor market impacts of EITC and welfare reform. This section discusses a subset of closely related papers, referring those interested in a more exhaustive summary to one of the many surveys of the literature.¹²

Most of the EITC literature has focused on single mothers, using Current Population Survey (CPS) data and variation from the federal EITC expansions in the 1980s and 1990s. These papers consider difference-in-differences approaches that rely on EITC variation by the presence and/or numbers of children. The early study by Eissa and Liebman (1996) finds sizable employment effects based on comparing single women with and without children before and after the 1986 EITC expansion. Meyer and Rosenbaum (2001) expand the time period to include the 1986, 1990 and 1993 tax reforms, and take a more structural approach based on modeling income taxes and welfare parameters from AFDC, Food Stamps, and Medicaid. They estimate that the EITC and other tax changes account for over 60 percent of the employment increase of single mothers between 1984-96. Hotz, Mullin, and Scholz (2006) use administrative panel data from California covering families on welfare during the period 1991-2000. Their empirical strategy exploits the differential EITC expansion for families with two or more children relative to families with one child after the 1993 reform. They find sizable employment effects. Gelber and Mitchell (2011) confirm the qualitative findings from these earlier studies using PSID data from 1975-2004. Hoynes and Patel (2018) provide an event study analysis of the 1993 reform. Their focus is on poverty effects, but their online appendix shows employment effects as well. Bastian (2020) estimates sizable employment effects of the 1975 EITC introduction based on comparing all women (single and married) with and without children.

Schanzenbach and Strain (2021) provide a comment on an earlier version of this paper (2019 working paper). They replicate findings from my paper and consider alternative specifications. Their preferred specification produces large effects of the EITC on extensive margin labor supply. They present only reduced-form estimates of employment effects, but I show that their estimates

¹²The EITC literature has been reviewed by Hotz and Scholz (2003), Eissa and Hoynes (2006), Meyer (2010) and Nichols and Rothstein (2015), while the welfare reform literature has been reviewed by Blank (2002) and Grogger and Karoly (2005).

correspond to very large elasticities (arguably outside the realm of the possible). These estimates are strong outliers in the distribution of estimates across a wide range of specifications.

A different approach to estimating labor supply responses to the EITC has been proposed by Chetty, Friedman, and Saez (2013). Based on the idea that responding to the EITC requires knowledge of the EITC, they estimate behavioral responses by comparing zip-codes that vary by EITC knowledge. Their proxy for knowledge is based on bunching by self-employed individuals at the first kink of the EITC. They back out extensive margin responses by comparing event studies of first child birth in high- and low-bunching areas. Women are more likely to continue working after child birth in high-bunching ("EITC") areas than in low-bunching ("no-EITC") areas. This approach yields an extensive margin elasticity of 0.19 at the average level of EITC knowledge between 2000-2005.¹³ Given their proxy for knowledge has increased significantly over time, their study suggests an average elasticity close to zero around the major EITC reforms in the early 1990s and before.

In the experimental literature, there are very few examples of EITC-style programs. Card and Hyslop (2005) study a temporary earnings subsidy to welfare recipients in Canada. Similar to the EITC, the transfer was conditional on working. But the program was otherwise different because it was time-limited (available for maximum of 3 years) and because eligibility required finding work within 12 months of random assignment, creating a strong short-term incentive for working to obtain the option value of future transfers. Such a scheme creates intertemporal substitution incentives and relates to the Frisch elasticity rather than the Hicksian steady state elasticity (see Chetty, Guren, Manoli, and Weber 2013). The closest we have to randomized evaluations of the EITC comes from the Paycheck Plus demonstration in New York City and Atlanta (Miller, Katz, Azurdia, Isen, Schultz, and Aloisi 2018; Yang, Bernardi, Metz, Miller, Katz, and Isen 2022). As discussed in the introduction, these trials find either modest effects (New York City) or zero effects (Atlanta) on employment, similar to the quasi-experimental findings in this paper.

The paper also contributes to the literature on welfare reform, including both observational and experimental studies (see Blank 2002; Grogger and Karoly 2005). Because the federal government required experimental evaluations of state waiver programs, there is a substantial body

¹³This empirical strategy relates to the literature estimating child penalties based on event studies of first child birth (Kleven, Landais, and Søgaard 2019; Kleven, Landais, Posch, Steinhauer, and Zweimüller 2019; Kleven 2023). In essence, the approach attributes the entire spatial correlation between child penalties and self-employed bunching to a causal effect of the EITC. The identification assumptions are quite strong, especially when considering that the child penalty literature finds very small effects of policies that create strong and salient extensive margin incentives such as maternity leave schemes and child care provision (see e.g., Kleven, Landais, Posch, Steinhauer, and Zweimüller 2022).

of experimental work assessing the impact of those programs. These trials generally bundled a range of treatments such as time limits, work requirements, welfare-to-work training, and financial incentives. The literature finds significant effects of welfare reform on extensive margin labor supply, consistent with the arguments in this paper. For example, Kline and Tartari (2016) study the extensive margin impacts of Connecticut's Jobs First Program, a randomized waiver demonstration that included time limits, work requirements, family caps, and earnings disregards. They find substantial effects on extensive margin labor supply.

There are relatively few studies that attempt to quantify the relative contribution of EITC and welfare reform to the employment increase of single mothers in the 1990s. Meyer and Rosenbaum (2001) estimate that the EITC accounts for 35% of the increase in employment between 1992 and 1996, with a more modest role for welfare reform. Similarly, Grogger (2003) estimates that the EITC explains 34% of the employment increase between 1993 and 1999, while welfare reform explains a much smaller share. Fang and Keane (2004) provide a detailed study of all the different policy parameters and macroeconomic variables that may have influenced the behavior of single mothers. Their estimates imply that the rise in employment between 1993 and 2002 can be explained in roughly equal proportions by the EITC, welfare reform, and the macroeconomy.

Of crucial importance to interpretation is the literature on EITC knowledge. In their study of extensive margin responses to the 1986 EITC expansion, Eissa and Liebman (1996) cite evidence from interviews conducted in 1993 showing "virtually no awareness" of the credit among potential recipients (Eissa and Liebman 1993; Olson and Davis 1994). A number of subsequent studies have documented the presence of substantial frictions related to the awareness, understanding, and claiming of the EITC (Romich and Weisner 2000; Smeeding, Phillips, and O'Connor 2000; Phillips 2001; Berube, Kim, Forman, and Burns 2002; Maag 2005; Kopczuk and Pop-Eleches 2007; Jones 2010; Chetty and Saez 2013; Mead 2014; Bhargava and Manoli 2015). For example, Maag (2005) reports that only 58% of low-income families "had heard about the EITC" in a nationally representative sample from 2002. What is more, the understanding of how the schedule is designed is even weaker. Among the families interviewed by Romich and Weisner (2000), most people had heard of the EITC, but virtually no one knew that they needed to earn a certain amount to maximize the credit. Chetty, Friedman, and Saez (2013) show that their proxy for knowledge — the degree of bunching by the self-employed at the first EITC kink — has been increasing over time and was extremely limited in the mid-nineties.

While these informational and psychological frictions are widely acknowledged, they have

been used mostly to explain the absence of *intensive* margin responses. The argument seems to be that extensive margin responses can be based solely on knowing about the existence of a tax refund, while intensive margin responses require detailed understanding of the schedule. This view is not consistent with economic theory, nor with basic intuition. Theoretical models predict that the extensive margin decision depends on taxes and transfers at the desired level of earnings, i.e. the intensive and extensive margin decisions are inter-dependent (see e.g., Kleven and Kreiner 2006; Eissa, Kleven, and Kreiner 2006, 2008). This inter-dependence is particularly strong for non-linear incentives like the EITC. In a model with fixed costs of working, the EITC is an incentive to start working around the refund-maximizing earnings range, not in any earnings range. Absent information about where the relevant earnings range is located and the size of the credit in that range, the extensive margin response is not very plausible.

3 Institutional Background and Data

3.1 The Earned Income Tax Credit

Since its inception in 1975, the Earned Income Tax Credit (EITC) has grown to become the largest cash transfer program in the United States. The EITC is a refundable tax credit and eligibility depends on having positive earned income. Figure 1 illustrates the federal EITC schedule in 2018 (Panel A) and the expansion of EITC generosity between 1968-2018 (Panel B). The credit amount is a nonlinear function of household earnings and the number of qualifying dependents (children).¹⁴ For each family size, the credit schedule features a phase-in range, a refund-maximizing plateau range, and a phase-out range. The generosity of the schedule is increasing in family size, with the largest possible credit for families with three or more children. Panel B shows the maximum possible credit for different family sizes over time.¹⁵ There have been five federal reforms of the EITC: the introduction in 1975 and the expansions enacted in 1986, 1990, 1993 and 2009.¹⁶ The largest changes are the 1975 EITC introduction to all families with children and the 1993 EITC expansion to families with two or more children.

¹⁴Specifically, a qualifying dependent is a relative who is under the age of 19 (24 for full-time students) or permanently disabled, and who resides with the tax filer for at least half the year.

¹⁵The maximum credit is shown in 2018 US Dollars. Table A.1 in the online appendix documents the full set of EITC parameters for each family size and each year between 1975-2018.

¹⁶These EITC changes were legislated as part of the Tax Reduction Act of 1975, the Tax Reform Act of 1986 (TRA86), the Omnibus Budget Reconciliation Acts of 1990 and 1993 (OBRA90 and OBRA93), and the American Recovery and Reinvestment Act of 2009 (ARRA).

Figure 2 illustrates the impact of these EITC expansions on the extensive margin incentives of single mothers. The figure shows the average tax rate on single women by number of children over the period 1968-2018. The tax rate includes state income taxes, federal income taxes, and federal payroll taxes.¹⁷ The calculations assume that women enter the labor market at the first kink of the federal EITC (in each year and for each number of children), i.e. where the EITC refund is maximized and the extensive margin incentive is the strongest. The average tax rate on single mothers has fallen enormously over this period, between 45 and 55 percentage points depending on the number of children. Almost all of this fall can be attributed to the five federal EITC reforms, and especially to the 1975, 1993, and 2009 reforms. The tax rate impact of the 1986 and 1990 reforms, on the other hand, is relatively modest.

In addition to the federal EITC, many states have introduced EITC supplements. A total of thirty states instituted EITC supplements between 1984-2018.¹⁸ Most state EITCs are specified as percentages of the federal EITC, implying that they have the exact same structure as the federal program. Table A.2 in the appendix provides details on all state EITC reforms, including their impact on the average tax rate of single mothers. While a number of state reforms did not change tax rates by very much, others introduced sizable changes (such as the reforms in Connecticut, Washington D.C., Michigan, New Jersey, and New York). The state reforms are useful for studying labor supply by introducing an additional source of variation: besides comparing single women with and without children (or with different numbers of children), they allow for comparing states with and without EITC supplements.

3.2 Welfare Reform

It would be impossible to assess the EITC without considering the confounding effects from the rest of the welfare system. In particular, the 1993 EITC expansion coincided with dramatic changes to traditional cash welfare. The Personal Responsibility and Work Opportunity Reconciliation Act of 1996 (PRWORA) implemented welfare reform at the national level, replacing Aid to Families with Dependent Children (AFDC) with the more restrictive Temporary Assistance for Needy Families (TANF). Importantly, this welfare reform did not represent a sudden and unanticipated departure from past policy. It was the culmination of state-led welfare reform efforts starting in the late 1980s and accelerating through the first part of the 1990s. These state reforms were imple-

¹⁷The calculations are based on NBER's tax simulation model (TAXSIM). Full details are provided in section C of the online appendix.

¹⁸Appendix Figure A.2 provides maps that illustrate the roll-out of state EITCs across the U.S.

mented under the heading of *welfare waivers*, federal approvals for states to change their welfare programs. I will highlight the key features of state and federal welfare reform below, but refer to Grogger and Karoly (2005) for an exhaustive description of the history and details of these policy changes.

Between 1992 and 1996, thirty-eight states received approvals for statewide legislation through waivers.¹⁹ There were six main types of waiver policies. *Termination time limits* introduced upper bounds on the length of time that any family could receive welfare benefits. *Work requirement time limits* imposed mandatory work requirements after a certain amount of time on benefits.²⁰ Building on the Job Opportunities and Basic Skills Training Program (JOBS) instituted in 1988, JOBS waivers strengthened the rules regarding participation in education, training, and job search activities. There were two such waivers. *JOBS exemptions* allowed states to eliminate or reduce the exemption of families with young children from the program. *JOBS sanctions* allowed states to eliminate or reduce benefit increases for existing AFDC recipients who had additional children. Finally, *earnings disregards* provided stronger financial incentives to work by disregarding earnings up to a level in the calculation of benefit claw back. Table A.3 in the appendix shows the approval and implementation dates of all statewide waivers.

The era of waiver-based reform culminated in national reform through PRWORA, signed into law in August 1996. This reform consolidated the AFDC and JOBS programs into the TANF program, which included all the key waiver elements just described. States had considerable latitude in designing their TANF programs under some federal guidelines. For example, it was a federal requirement that states impose a time limit of no more than 60 months, but they were free to choose stricter limits and many did. States without any time limit had to introduce one. As a result, TANF extended the waiver-type elements that had already been implemented in a number of states to the remaining states.

Welfare reform did not primarily change statutory benefit levels, but imposed much harsher constraints on receiving those benefits by strengthening ordeals and enforcement. This increased the incentives for single mothers to move from welfare into work and mechanically kicked some of them off welfare. Importantly, the treatment intensity was not uniform across single mothers, but increased in the number of children. The main reason is that, prior to welfare reform, AFDC

¹⁹See Council of Economic Advisors (1997, 1999) and Department of Health and Human Services (1999).

²⁰Henceforth, I refer to termination time limits simply an "time limits" and to work requirement time limits as "work requirements".

participation rates were sharply increasing in family size. This is shown in Figure A.3 of the online appendix. Therefore, as welfare reform restricted benefit access across all family types, the implications were stronger for larger families as they were more reliant on benefits to begin with. This *strictly* increasing relationship between number of children and treatment intensity separates welfare reform from EITC reform. The 1993 EITC expansion was larger for families with 2 children than for families with 1 child (like welfare treatment), but the same for families with 3+ children as for families with 2 children (unlike welfare treatment). In this paper, I use family size (along with other demographic variables that predict treatment intensity) to tease apart the impacts of EITC and welfare reform in the 1990s.

3.3 Data

The analysis is based on the Current Population Survey (CPS). I combine data from the basic monthly files and from the Annual Social and Economic Supplement (ASEC), or "March files". Merging the monthly and March files give much more data than using the March files alone, which is what most previous papers in the literature have done.²¹ I restrict the dataset to include the monthly files from 1989-2019 and the March files from 1968-2019. Even though the monthly files go back to 1976, they do not allow for accurately identifying the presence and number of children prior to 1989. I focus on the sample of single women aged 20-50.²² These restrictions result in a sample of 4,809,195 individual-month observations across survey years 1968-2019. Appendix B provides a detailed description of the CPS data.

Extensive Margin Measures: The CPS allows for four different ways of measuring labor supply at the extensive margin: employment at the weekly or annual levels, and labor force participation at the weekly or annual levels. As shown in Figure A.5 of the appendix, the four extensive margin measures are highly correlated over time. I use weekly employment as the baseline outcome, but reproduce all the main analyses using the other outcomes as well. Given most of the existing literature has focused on annual employment, it is worth outlining the conceptual trade-offs in the choosing the extensive margin measure.

First, the difference between employment and labor force participation is that the latter includes unemployed people who are actively searching for work or only temporarily laid off. While

²¹The main advantage of the March supplement is that it contains detailed information on annual income variables during the previous year. I use the March files alone for analyses that requires annual income information.

²²The definition of "single" includes never married, separated, divorced, and widowed.

employment is more relevant for welfare calculations, labor force participation has the advantage of not moving with transitions between employment and unemployment. This makes participation less vulnerable to the confounding effects of the business cycle. Second, the difference between annual and weekly measures is that the latter, besides capturing extensive margin responses, may also capture intensive margin responses (working more or less weeks over the year, conditional on working). Given the existing literature finds small or zero intensive margin responses to the EITC, this difference between annual and weekly measures seems quantitatively trivial.²³ That is, given the evidence on intensive margin responses *and* absent any statistical differences between the two measures, we would expect to get the same results from both.

However, there are important statistical advantages from using weekly measures. First, the weekly measures are available in both the monthly and the March files, while the annual measures are available only in the March files. Weekly measures therefore give much more statistical power. Second, questions about work activities during the previous week likely involve less measurement error than questions about work or earnings during the previous year. Third, the weekly measures largely avoid issues with nonresponses in the CPS. There are very few nonresponses in the weekly labor market and demographic variables, while there is significant nonresponse in the annual income variables in the March files. As a result, a significant fraction of earnings observations in the CPS are based on imputations.²⁴ For these three reasons, weekly measures are preferred in terms of measurement and statistical power.

Comparison Groups: The empirical design exploits the two key determinants of EITC eligibility: children and earnings. The existing literature has compared *all* single women with and without children or *low-educated* single women with and without children. I use all single women as my baseline sample, but reproduce the main results for low-educated single women in the appendix. The goal of using education to restrict the sample is to capture families with low earnings capacity and therefore stronger EITC treatment. However, cutting the sample by education level raises two issues: (i) proxying for earnings capacity using only education is relatively crude given the many other demographic variables that affect earnings, and (ii) using a fixed education cutoff

²³Chetty, Friedman, and Saez (2013) do find evidence of intensive margin responses to the EITC, but the average response is small due to offsetting effects from the phase-in and phase-out regions.

²⁴These imputations are described in Appendix B. Bollinger, Hirsch, Hokayem, and Ziliak (2019) provides a detailed analysis of the issue.

introduces selection bias due to the increase in education over time.²⁵ To avoid both of these issues, I take an approach based on fixed quantiles of predicted earnings. Specifically, I predict earnings conditional on working based on the following regression implemented on the sample of single women with positive earnings:

$$Y_i = \alpha_a + \beta_n + \gamma_y + \delta_e + \zeta_r + \eta_s + \nu_i, \tag{1}$$

where Y_i is the earnings of woman *i* and the right-hand side includes fixed effects for the age of the woman *a* (6 categories), the number of children *n* (7 categories), the age of the youngest child *y* (7 categories), education *e* (4 categories), race *r* (4 categories), and state *s*.²⁶ The estimated coefficients are used to predict earnings \hat{Y}_i in the full sample of single women (workers and nonworkers), providing a virtually continuous measure of EITC treatment intensity. I then investigate the impact of the EITC in different quantiles of the predicted earnings distribution. These quantiles (such as the bottom quartile or bottom half of the distribution) are selected within each year to ensure that the sample selection is consistent through time.

Descriptive Statistics: Table 1 provides descriptive statistics for single women with children (columns (1)-(3)) and single women without children (columns (4)-(6)). In each group, outcomes are shown for the full sample and for subsamples with low education (high school degree or less) or low predicted earnings (bottom quartile of the distribution).²⁷ For single women with children, average earnings equal \$22,186 in the full sample, \$15,313 in the low-education sample, and \$10,922 in the low-earnings sample. Contrasting women in the low-education and low-earnings samples, we see that the latter are almost as low educated, younger, more African-American, and that they have more and younger children. Appendix Figure A.6 shows the earnings distribution of single mothers in the three samples, normalizing earnings by the first kink of the EITC in each year (so that the kink is located at zero). In all samples, the vast majority of single mothers have earnings within the EITC range: 73.3% in the full sample, 83.2% in the low-education

²⁵Consider the typical education cutoff used in the literature: high school degree or less. At the beginning of my sample period, 87% (59%) of single women with children (without children) were low-educated by this definition. At the end of my sample period, only 44% (27%) of single women with children (without children) were low-educated. Hence, using a fixed education cutoff is associated with substantial changes in sample selection that differ between treatment and control groups.

²⁶The binning of these dummy variables is as follows: age of woman (20-24, 25-29, 30-34, 35-39, 40-44, 45-50), number of children (0, 1, 2, 3, 4, 5, 6+), age of youngest child (0-1, 2-3, 4-6, 7-9, 10-13, 14-17, 18+), education (below high school, high school degree, some college, college degree and above), and race (white, black, Asian, and other).

²⁷All samples include single women in the age group 20-50 and pool data from survey years 1968-2019. Earnings are reported in 2018 USD.

sample, and 93.6% in the low-earnings sample. Results for the full sample can be interpreted as intention-to-treat (ITT) estimates, while results for the low-earnings sample are close to treatment-on-the-treated (TOT) estimates.

4 EITC and the Extensive Margin: The Long View

I start from a long-run perspective, documenting the evolution of extensive margin labor supply for single women over the last fifty years. Panel A of Figure 3 compares the labor force participation rates of single women with and without children between 1968-2018. Given EITC eligibility depends on children, we can think of these series as treatment and control groups.²⁸ The patterns are striking. Over half a century, all of the action in the participation rate of single mothers relative to that of single childless women took place during a single spell in the mid-late 1990s. Outside this spell, the two groups have evolved in parallel. At the beginning of the period, in 1968, the gap in labor force participation between the two groups was equal to 14.3 percentage points. A quarter of a century later, in the early 1990s, the gap was about the same. Then the labor force participation of single mothers rose dramatically in the mid-late 1990s, closing the entire gap in just a few years. After this, the two groups went back to parallel trends — but now at the same levels — and have stayed that way to this day. The graph highlights how exceptional the 1990s were in the history of the U.S. labor market.

What explains this long-run evolution? To consider the role of the EITC and other factors, Panel B of Figure 3 compares the long-run series to the timing of the five federal EITC reforms and the confounders from welfare reform and the macroeconomy. The EITC's claim to large extensive margin effects relies on two correlated events: the EITC expansion enacted in 1993 and the ensuing increase in the participation rate of single mothers. At the same time, the figure highlights a major puzzle. If EITC reform drove the extensive margin increases in the 1990s, then why do we not see any such effect around the other reforms? The three EITC reforms prior to 1993 did not close any of the gap between single women with and without children, nor did the EITC expansion in 2009. As discussed above, while the 1993 tax reform introduced very large tax cuts on single mothers, so did the 1975 and 2009 tax reforms.²⁹

Any compelling EITC narrative must reconcile the starkly different patterns observed in the

²⁸Childless families could not receive any EITC until 1993 and only a very modest credit thereafter (see Figure 1).

²⁹Looking closely at the long-run series, the participation gap narrows marginally between 1976-78 — after the 1975 EITC introduction — although it widens again later. As I show below, there is no such short-term narrowing of the employment gap in the 1970s.

1990s and elsewhere. The missing link could be the two confounders emphasized here. First, there was welfare reform: waiver-based state reforms between 1992-96, culminating in national reform in 1996. Second, there was the macroeconomy: the economy was booming during the Clinton era, with the national unemployment rate falling steadily between 1992-2000. Why would the macroeconomy impact single women with and without children differently? Apart from the 1990s, there is no clear correlation between unemployment and the participation gap between the two groups. This is not inconsistent with the business cycle having an effect in the 1990s, however, because of the interaction between the business cycle and welfare reform. As single mothers get pushed off welfare, their ability to find work will depend on the tightness of the labor market around the time of the reform. Moreover, any short-run interaction between welfare reform and the business cycle may persist in the long run due to employment hysteresis (e.g., Yagan 2019).

The long-run series in Figure 3 show weekly participation rates in the sample of all single women. Figures A.7-A.9 in the appendix replicate the analysis for all four extensive margin measures (participation and employment at the weekly and annual levels) and three samples (all single women, low-educated single women, and single women with low predicted earnings).³⁰ The figures show that the qualitative patterns are robust to the choice of extensive margin measure and sample.³¹ In the low-educated sample, there is some shrinking of the extensive margin gap between single women with and without children in the decades leading up to the 1993 reform. However, as discussed above, cutting the sample by education introduces sample selection bias due to the large increase in education levels over time, making it more appropriate to cut the sample by within-year quantiles of predicted earnings. In the low-earnings sample, there is a *widening* of the gap between single women with and without children over time. Hence, zooming in on a sample more strongly treated by the EITC does not reconcile the contrast between the 1990s and other periods. If anything, it makes the contrast even sharper.

Further insight can be gained by splitting the sample of single mothers into different family sizes. Figure 4 shows the long-run evolution in labor force participation for single women with zero, one, two, and three or more children. The broad pattern is the same as before: the different groups have trended similarly over half a century, except for the mid-late 1990s. However, while

³⁰See section 3.3 for details. Low-educated single women are defined as those with a high school degree or less. Single women with low predicted earnings are defined as those below the median of the within-year distribution of predicted earnings. The annual participation series is shorter than the other series because this measure was not recorded in CPS data until the 1976 March files (pertaining to calendar year 1975).

³¹In general, the employment series are bumpier than the participation series. This is because employment is more sensitive to the business cycle as it is directly affected by movements between employment and unemployment.

all groups of single mothers increased labor force participation in the 1990s, the magnitude is strongly increasing in family size. Consider those with three or more children: their participation rate increased by a staggering 23 percentage points over six years, about twice as much as for those with two children. This difference is puzzling under the EITC narrative, because the tax credit expansion was the same for those with two and three children. It is of course possible that larger families have larger extensive margin elasticities and therefore respond more strongly to the same incentive, but this line of reasoning leads to other puzzles: if single women with three or more children are much more elastic, then why do they not increase participation after the 2009 EITC expansion (targeted specifically to them) or after the 1986 and 1990 expansions?³²

To conclude, the long-run evidence suggests that the dramatic changes for large female-headed families in the mid-late 1990s were driven by factors that were unique to that time period *and* strictly increasing in family size. Welfare reform satisfies both; EITC reform satisfies neither.

5 The Impact of Federal EITC Reforms

5.1 Event Studies of Federal EITC Reforms

This section presents event studies of the five federal EITC reforms in 1975, 1986, 1990, 1993, and 2009.³³ The baseline specification uses weekly employment as the outcome and all single women as the sample, but alternative outcomes and samples will be considered below. The empirical strategy is a difference-in-differences (DiD) design comparing single women with and without children around each reform.³⁴ The event study specification is the following:

$$P_{imt} = \sum_{j} \alpha_{j} \cdot Y ear_{j=t} + \beta \cdot Kids_{i} + \sum_{j \neq -1} \gamma_{j} \cdot Y ear_{j=t} \cdot Kids_{i}$$
(2)
+ $X_{i}\phi + \eta \cdot U_{st} + \theta \cdot U_{st} \cdot Kids_{i} + \nu_{imt},$

³²Figures A.10-A.12 in the appendix replicate the long-run series by family size for the different extensive margin measures and samples. The stylized patterns are robust across outcomes and samples.

³³Because the CPS monthly files can be used only from 1989 (see section 3.3), to ensure that each reform is analyzed using consistent data through time, I use March files alone for reforms that occurred before 1993 and March and monthly files combined for reforms that occurred from 1993 onwards. The same split is made in the analysis of state EITC reforms in the next section.

³⁴The 2009 reform is an exception. Here I compare single women with three or more children to single women without children, because this EITC expansion was targeted specifically to families with 3+ children. Alternative comparison groups (such as 3+ vs 2 children) give similar results, but because single women with 1 or 2 children were strongly treated by non-EITC aspects of the 2009 reform (see Figure 2), the comparison groups used here are econometrically more convincing.

where P_{imt} is an indicator for individual *i* working in month *m* of year *t*. The right-hand side includes dummies for each year, a dummy for having kids, the interaction between year and kids dummies, along with controls for demographics and the business cycle. By omitting the year just before the reform (indexed as -1) in the interaction term, the DiD coefficients γ_t can be interpreted as the extensive margin effect in year *t* relative to the pre-reform year. The vector of demographic controls X_i includes dummies for the age of the woman (6 categories), the age of the youngest child (7 categories), education (4 categories), race (4 categories), and state.³⁵ To adjust for business cycle confounders, the specification adds controls for the aggregate unemployment rate in state *s* and year *t*, U_{st} , interacted with kids. Specifically, U_{st} is a demeaned unemployment rate in state *s* over the event study window. This captures the idea that business cycle variation reflects movements in unemployment relative to its "structural" level.

In addition to event studies of each reform separately, I present stacked event studies of several reforms combined. Interacting the specification in (2) with indicators for each reform episode k, I obtain DiD coefficients γ_t^k for event year t and reform k. Stacked DiD coefficients are calculated as $\bar{\gamma}_t = \sum_{k=1}^K \gamma_t^k / K$, where K is the number of reforms included. These coefficients give the average effect across different reforms, weighing each reform equally. In the stacked event studies, I always exclude the 1990 reform as a separate reform episode due to its close proximity to the 1986 reform (given the event study window used, any effect of the 1990 reform will be captured by the 1986-reform coefficients). I show results from two versions of the stacked event study, one that excludes the confounded 1993 reform and one that includes it.

The first set of results is presented in Figure 5. The figure plots DiD coefficients $\hat{\gamma}_t$ for the 1975 reform (Panel A), the 1986 and 1990 reforms (Panel B), the 1993 reform (Panel C), the 2009 reform (Panel D), federal reforms stacked excluding the 1993 reform (Panel E), and federal reforms stacked including the 1993 reform (Panel F). The panels also report the average three-year effect of each reform (with its standard error in parenthesis). The 3-year window is chosen to avoid overlap between reforms.

The results confirm the main insight from the descriptive long-run analysis presented above, namely that the patterns around the 1993 EITC expansion are exceptional. Taken at face value, the 1993 event study looks compelling: pre-trends trends are roughly parallel, employment starts increasing in the treatment group relative to the control group after the reform, and a large and

³⁵The binning of these dummies is the same as for the earnings regression (1) described above.

statistically significant effect builds up over time. The average three-year effect (1994-96) on the employment rate equals 1.29pp. This precedes the passage of national welfare reform in August 1996, but coincides with the period of waiver-based state welfare reforms in 1992-96. The effect grows much larger in the late 1990s, following the passage of national welfare reform. No other EITC reform is associated with such patterns. The event studies are either flat around the reform implementations (1975 and 1986) or falling (1990 and 2009). As a result, the stacked event studies are also flat and this holds even when including the confounded 1993 reform. The average 3-year effect is negative and statistically insignificant in both stacked event studies.

Table 2 presents estimates from different specifications and reforms. Results are shown for weekly employment and annual employment, and for specifications without controls, with only demographic controls, and with both demographic and unemployment controls. The baseline results from the preceding figure correspond to those in column (3) of the table. All effects are averages over three post-reform years. Besides showing reduced-form effects of each tax reform on employment rates, the table converts the effects into extensive margin elasticities with respect to the net-of-tax rate.

The extensive margin elasticity is defined as

$$\varepsilon \equiv \frac{\Delta P/P}{\Delta \left(1-\tau\right) / \left(1-\tau\right)},\tag{3}$$

where $\Delta P/P$ is the percentage effect of the reform on the employment rate and $\Delta (1 - \tau) / (1 - \tau)$ is the percentage effect of the reform on the average net-of-tax rate.³⁶ Both the numerator and denominator of the elasticity are measured based on DiDs comparing treatment and control groups. To simplify the calculation of elasticities across many reforms and over a long time period, I use the average tax rate series presented in Figure 2 to get the denominator. This implies that two approximations are made. First, while the calculations account for the entire tax system, they ignore the implicit tax from the welfare system. This overstates the *baseline* net-of-tax rate $1 - \tau$. Second, the calculations assume that single women enter the labor market at the first kink of the federal EITC, i.e. where the EITC refund is maximized. This overstates the *change* in the net-of-tax rate $\Delta (1 - \tau)$ induced by EITC reform, because some workers enter at higher earnings levels where the EITC expansions were weaker. Therefore, the two approximations have offsetting effects on

³⁶This elasticity definition corresponds to the literature on optimal taxation and welfare measurement (see e.g., Saez 2002; Kleven and Kreiner 2005; Eissa, Kleven, and Kreiner 2006, 2008).

the denominator of the elasticity.³⁷ Full details of the tax simulations and elasticity calculations are provided in Appendix C.

Table 2 confirms the main insight from the event study graphs. Except for the 1993 reform, none of the estimates are positive *and* statistically significant at the conventional 5% level. Across all 30 estimates from outside the 1993 reform, two are positive and significant at the 10% level, while all others are either negative or statistically insignificant even at the 10% level. Two additional points are worth highlighting. First, while the estimates for the 1993 reform are sizable across all specifications, they decline substantially as controls are added. For example, the elasticity of weekly employment goes from 0.53 without any controls to 0.22 with controls for demographics and the business cycle. This is a concern for causal inference, because there is no guarantee that these controls are enough. For example, the state-level unemployment controls included in (2) do not address national business cycle effects present in all states. Second, there is large variation in the elasticity estimates based on the 1986 reform, from -0.48 to +0.75 across specifications. The reason is that the 1986 reform introduced only small tax rate changes for single mothers, implying that even modest bias in the reduced-form estimates translates into large bias in the elasticity estimates.

In the following sections, I delve deeper into the implications of specification choice and the connection between my results and the prior literature.

5.2 Treatment Effects by Predicted Earnings

Estimations based on the full sample of single women do not account for the fact that the EITC program is means-tested. This may attenuate the effects as some single mothers are ineligible due to having earnings above the EITC exhaustion point. As shown in Appendix Figure A.6, most but not all single mothers have earnings within the EITC range. This is consistent with the possibility of some attenuation in the full sample. To put it differently, estimations for the full sample give intention-to-treat (ITT) effects rather than treatment-on-the-treated (TOT) effects. To see if this distinction matters here, I investigate responses among single women with low predicted earnings, estimated using equation (1) described above.

Figure 6 shows event studies for single women below the median of predicted earnings, while Table 3 shows average three-year effects across different specifications for the same sample. Apart

³⁷Results for the 1993 reform (available upon request) show that, in practice, the two approximations roughly cancel out, so that the elasticities presented here provide good approximations of the exact elasticities.

from the sample, these exhibits are constructed exactly as those presented above. Restricting the sample to single mothers more strongly treated by the EITC does not change the main insights. The effects of the 1993 reform (which includes the confounding effects of welfare reform) are larger in this sample, but there are still no clear effects around any of the other reforms. The stacked event studies are virtually flat around the reform experiments, with a slightly negative and statistically insignificant effect during the first three post-reform years.

Is the median sample split too coarse to capture the subset of women who are responding to the EITC? I investigate this possibility in Figure 7, which plots three-year effects of federal EITC reforms by decile of predicted earnings. The estimates are based on stacked DiD specifications including either all reforms (red dots) or all reforms apart from 1993 (black dots). Panel A shows estimates from a specification without controls, while Panel B adds controls for demographics and state-level unemployment rates. As can be seen from the figure, the profile of DiD coefficients is flat around zero across the distribution of predicted earnings. This holds with and without controls and whether or not the 1993 reform is included. There is no earnings decile where the effect is positive and statistically significant. If the EITC had an effect on extensive margin labor supply, we would expect the estimates to be positive in the lower deciles of predicted earnings and zero in the upper deciles.³⁸

5.3 1975 EITC Introduction: Single vs Married Women

A recent paper by Bastian (2020) studies the 1975 EITC introduction, estimating that the reform increased maternal employment by about 6%. Bastian (2020) considers the effect on *all* mothers, single and married, while this paper focuses on the effect on single mothers alone. This section includes married women in the analysis, investigating if they strengthen the case for extensive margin reponses to the 1975 reform. When specifying an empirical model to estimate behavioral responses among married women, it is critical to account for the incentives created by the EITC's reliance on household income to determine eligibility. In fact, the EITC creates negative extensive margin incentives on most married women (Eissa and Hoynes 2004).

Married women with spousal earnings above the first EITC kink were either untreated or negatively treated by the introduction of the program. Their entry would trigger no tax credit if spousal

³⁸Figures A.13-A.16 in the online appendix show that these insights are robust to the extensive margin outcome used. The annual measures feature a statistically significant effect in decile 4 of predicted earnings, but no effect in any other decile. Given the large number of estimates in these figures (160 estimates), it is not surprising that some are statistically significant.

earnings were above the EITC exhaustion point and otherwise reduce the size of the credit.³⁹ Even married women with spousal earnings in the phase-in range were mostly negatively treated because, unless spousal earnings were tiny, their entry would take the family over the hump and reduce the size of the credit. This motivates a specification in which women are considered positively treated if they are either single or married with spousal earnings of zero (henceforth "treated" women). Otherwise, they are considered negatively treated or untreated (henceforth "untreated" women). Figure A.17 in the appendix presents results from such a specification. The figure is based on a triple-differences version of equation (2), adding a treatment dummy (and its interactions with year and kids) equal to one if a woman is either single or married with spousal earnings of zero.⁴⁰ The figure shows difference-in-differences series for both treated and untreated women (comparing those with and without children in each group) and reports triple-differences estimates of the average post-reform effect based on comparing the two difference-in-differences.⁴¹ Results are provided for both weekly and annual employment.

The figure shows that treated and untreated women track each other over time, consistent with no effect of the EITC on employment. The triple-differences estimate of the average effect is marginally negative and statistically insignificant.⁴² The figure also considers a specification in which single women are compared to all married women, regardless of spousal earnings. The results are based on the same triple-differences specification, but where the treatment dummy equals one for single women and zero for married women. The event study for single women is flat around zero, consistent with the findings in the previous sections. The event study for married women — most of whom are untreated or negatively treated by the EITC — shows employment increases in the post-reform period relative to the pre-reform period. As a result, the triple-differences estimates are strongly negative.

To summarize, including married women in the analysis and accounting for the incentives created by the jointness of the US income tax does not strengthen the case for extensive margin

³⁹Note that the original EITC schedule did not have a plateau range; the phase-out of the credit began immediately after the first kink. See Appendix Table A.1 for information on the federal EITC parameters over time.

⁴⁰The details of this specification are described in Appendix E.

⁴¹This empirical strategy is different from Bastian (2020). His event study evidence is based on a difference-indifferences specification that compares women with and without children in two alternative samples: the full sample of single and married women or a subsample that excludes married women with spousal earnings above the EITC exhaustion point. Both of these samples include married women who are either unaffected or negatively affected by the EITC in the treatment group.

⁴²Changing the cutoff for spousal earnings does not change this conclusion. Even if we include all married women with spousal earnings below the first EITC kink in the treatment group (which would be going too far), the triple-differences specification implies no effect of the EITC.

responses to the 1975 reform. If anything, the patterns for married women weaken the case. Untreated married women increased employment by at least as much as treated women following the reform, implying triple-differences estimates that are zero or negative.

5.4 Model Uncertainty

Estimating causal EITC impacts is challenging because of imperfections in the empirical design and survey data used. This creates uncertainty about how to specify an empirical model that avoids bias from confounding shocks and measurement error. The preceding sections considered the implications of different estimation samples, extensive margin measures, and control variables, arguing that the main insights are robust to changing the model in these dimensions. However, the specifications presented constitute just a small subset of the possible specifications. In this section, I take a more exhaustive approach to model uncertainty by presenting estimates from a wide range of specifications that vary the estimation sample, extensive margin outcome, and controls. Specifically, I consider all possible permutations of four samples (all single women, low-educated single women, single women below the median of predicted earnings, and single women below the 25th percentile predicted earnings), four outcomes (employment and participation at the weekly and annual levels), and four specifications of controls (no controls, basic demographic controls, rich demographic controls, and rich demographic controls plus unemployment controls).⁴³ With the seven reform experiments considered here (five federal reforms and two stacked specifications), this amounts to a total of 432 specifications.

The online appendix provides event studies for all 432 specifications.⁴⁴ Figure 8 shows the distribution of estimates obtained from these event study specifications. These are estimates of average three-year effects of each reform experiment. Panel A shows reduced-form effects on employment or participation rates in percentage points, while Panel B shows the implied elasticities with respect to the average net-of-tax rate. The gray bars depict the estimate distribution without the 1993 reform, while the blue bars depict the estimate distribution for the 1993 reform. These two distributions are very different. The non-1993 distribution is symmetric around zero, with a mean reduced-form effect of -0.32pp and a mean elasticity of -0.04. The 1993 distribution is shifted to the right, with a mean reduced-form effect of 4.02pp and a mean elasticity of 0.63. These distributions

⁴³The basic demographic controls include dummies for the age of the woman, the age of the youngest child, and education. The rich demographic controls add dummies for race and state. The binning of these dummy variables is the same as in equation (2) described above.

⁴⁴See Figures A.18-A.81.

cannot both be right unless the underlying tax elasticities were somehow much larger in the midlate 1990s than in any other period. A more plausible interpretation is that the 1993 distribution is upward biased due to the confounding effects of welfare reform and the macroeconomy.

Another way of illustrating model uncertainty is by way of a specification curve analysis as presented in Figure 9. This figure shows the curve of elasticity estimates, ranked from low to high, across all combinations of reform experiments and specifications. Leaving out the confounded 1993 reform, the figure includes 304 different elasticity estimates (black dots) and their 95% confidence intervals (gray bars). The underlying specification for each estimate is indicated below the graph. The striking finding is that less than a dozen specifications (3.6% of all specifications) yield estimates that are positive *and* statistically significant. Arguing that the EITC has sizable effects therefore requires taking a stand on specifications that are strong outliers in the distribution of estimates. Specifically, the only way to get large and statistically significant elasticities is to focus on the 1986 tax reform and annual employment (but *not* annual participation), combined with certain samples and specification of controls. These outlier estimates have large confidence intervals, preventing us from ruling out elasticities anywhere between marginally above zero and around 2.

Relying on the 1986 reform to obtain large elasticities is not convincing, because this reform was small in terms of the reduction in the average tax rate on single mothers (see Figure 2). Estimating elasticities from small tax variation raises two issues: (i) it is hard to accurately detect the response to a small incentive in noisy data, and (ii) any small bias in the reduced-form estimate (the numerator of the elasticity) translates into large bias in the elasticity estimate (due to the small denominator). It is therefore concerning that the few large and statistically significant elasticity estimates in Figure 9 come from the 1986 reform.

To conclude, considering hundreds of empirical specifications, the estimated impacts of EITC reform are concentrated around zero and are statistically insignificant. Outside the confounded 1993 reform, only a tiny fraction of specifications yield estimates that are both positive and significant. Of course, the analysis is based on a selected set of specifications (albeit a very large set) and others could be added, including alternative control variables or the type of triple-differences specifications considered in the previous section. Nevertheless, the analysis clearly demonstrates that the EITC consensus does not follow from any robust feature of the data.

5.5 Comparison with Existing Estimates

It is useful to compare the distribution of estimates across specifications to existing estimates in the literature. Therefore, Figure 10 compares the distribution of elasticities from Figure 8B (leaving out the confounded 1993 reform) to elasticities implied by the reduced-form estimates in Eissa and Liebman (1996) and Schanzenbach and Strain (2021). These two papers are chosen for several reasons: they are representative of the consensus in the EITC literature, they consider other reforms than 1993, and they provide reduced-form estimates that are easily converted into elasticities.⁴⁵ I use estimates from the preferred specifications in these papers, both of which consider effects on annual employment for either all single mothers or low-educated single mothers.⁴⁶ Full details on the conversion of these reduced-form estimates into elasticities are provided in Table A.4 of the online appendix.

In Figure 10, estimates based on Eissa and Liebman (1996) are demarcated by black vertical lines, while estimates based on Schanzenbach and Strain (2021) are demarcated by blue vertical lines. Apart from the 2009 reform (the two lines furthest to the left), their estimates are strong positive outliers in the distribution of estimates. The elasticities range from 0.30 to 1.66, all in the upper tail of the distribution. The largest estimates are those based on the 1986 reform for the sample of low-educated single women, an elasticity of 1.34 in Eissa and Liebman (1996) and 1.66 in Schanzenbach and Strain (2021). Elasticities of such magnitudes seem implausible. The reason why the 1986 tax reform can yield such large elasticity estimates is that this reform introduced only small changes in the average tax rates of single mothers, implying that even modest bias in the reduced-form estimates blows up to large bias in the elasticity estimates. As discussed above, it is challenging to accurately estimate elasticities based on small identifying variation in tax rates.⁴⁷

⁴⁵These and other papers in the EITC literature do not convert their reduced-form estimates into elasticities. Elasticity calculations are useful for several reasons, including the fact that they provide a normalized measure of responsiveness that can be compared across reforms and allows for assessing the plausibility of magnitudes (see Kleven 2021).

⁴⁶For Eissa and Liebman (1996), I use their estimates reported in Table III (column 5) for all single women and Table IV (columns 1-2) for low-educated single women. Their paper shows effects on single mothers with less than a high school degree and with a high school degree separately. To ensure comparability with the low-education sample considered here, I calculate a weighted average of their estimates using the pre-reform fractions of low-educated single mothers with less than high school and high school, respectively. Eissa and Liebman (1996) consider four-year effects of the 1986 reform (comparing 1984-1986 to 1988-1990). For Schanzenbach and Strain (2021), I use their estimates reported in Table 3 (column 2). These correspond to the estimates presented here: average three-year effects in the sample of either all single women or low-educated single women, defined as those with high school degree or less. I focus on their estimates for the 1975, 1986, 1990, and 2009 reforms. Both papers provide estimates of the reduced-form impact on the employment rate, ΔP . When calculating the percentage effects on the employment rate $\Delta P/P$ and on the average net-of-tax rate $\Delta (1 - \tau) / (1 - \tau)$, I replicate as closely as possible the estimations samples described in each paper. The elasticity conversions are done consistently with the elasticities presented above, i.e. using the average tax rate series in Figure 2 in which single women are assumed to enter the labor market at the first kink of the EITC.

 $^{^{47}}$ Chetty, Guren, Manoli, and Weber (2013) convert the Eissa-Liebman estimate for all single mothers into an elasticity

6 The Impact of State EITC Reforms

6.1 Event Studies of State EITC Reforms

In this section I consider a different source of variation: state EITC reforms. A total of 27 states have introduced and maintained an EITC supplement for a sufficiently long period of time to conduct an empirical study.⁴⁸ The introduction of state EITC supplements provides useful variation by allowing for comparing single mothers in states with and without an EITC, thus being immune to confounders that impact single mothers as a whole. State EITC reforms are analyzed using a synthetic control approach: for each state with an EITC supplement, I create a synthetic control state from the pool of states that never had an EITC supplement. I then run a stacked event study comparing treatment and synthetic control states around state EITC introductions.

Not all of the state EITC reforms provide useful quasi-experiments. I exclude three states as their EITC introduction was offset by other state tax changes, thereby *increasing* the total average tax rate on single women with children relative to those without.⁴⁹ I exclude another six states due to small sample sizes, less than 100 single mothers per year around the reforms.⁵⁰ Including these reforms in the analysis adds considerable noise in the estimates. These restrictions leave a total of 18 state EITC reforms. Some of these reforms were small, while others were quite large. I therefore compare specifications based on all 18 reforms to specifications that focus on the 10 largest reforms as measured by the reform-induced change in the state average tax rate on single women with children relative to those without.⁵¹ As can be seen in Table A.2, these reforms reduced the average tax rate on treatments relative to controls by between 3.05pp and 8.32pp. These tax calculations include all state income taxes (not just the EITC) and assume that women enter the labor market at the first kink of the federal EITC.

To run the synthetic control analysis, the CPS data is collapsed to state-by-year observations for

of 0.30, much smaller than the elasticity of 0.71 in Table A.4. This is despite the fact that they use a larger reduced-form estimate (taken from column 4 of Table III in Eissa and Liebman 1996). The reason for their smaller elasticity is that they scale the reduced-form effect using a much larger denominator, including the total change in net earnings from taxes, welfare benefits, and Medicaid between 1984-1990. This gives a denominator of 13.2% instead of the 4.2% used here. Using the total incentive change from taxes, welfare benefits, and Medicaid is problematic, because the research design of Eissa and Liebman (1996) is supposed to identify the causal impact of the 1986 Tax Reform Act alone. In other words, such an elasticity calculation can only be correct if the Eissa-Liebman estimates are fully confounded by welfare and Medicaid changes.

⁴⁸Table A.2 lists 30 states with an EITC supplement. But the state of Washington never funded or paid out the credit, while Hawaii and South Carolina introduced their supplements only in 2018.

⁴⁹These states are Indiana, Ohio, and Oregon. See Table A.2.

⁵⁰These states are Iowa, Maryland, Minnesota, Rhode Island, Vermont, and Wisconsin.

⁵¹These are the EITC reforms in Colorado, Connecticut, District of Columbia, Kansas, Massachusetts, Michigan, Nebraska, New Jersey, New Mexico, and New York.

single women with and without children separately. For each treatment state, a synthetic control state is constructed (from among those without an EITC supplement) by matching on the level of the outcome variable in each of the five pre-reform years.⁵² I consider two empirical approaches. The first is a difference-in-differences approach that compares treatment and control states in the sample of single mothers alone. The second is a triple-differences approach that compares treatment and control states for single women with children relative to those without. The full details of these specifications are provided in Appendix F.

Figure 11 shows four event studies: difference-in-differences vs triple-differences (left vs right panels) and all reforms vs the ten largest reforms (top vs bottom panels). As above, I start by considering the sample of all single women and using weekly employment as the extensive margin measure. The figure shows clearly that there is no effect of state EITC supplements in any of the four event studies. The treatment and control states track each other in the pre-reform years (by construction) and continue to do so in the post-reform years. This is despite the fact that the tax variation is sizable, a reduction in the average tax rate of about 6pp across the ten largest reforms.

6.2 Model Uncertainty

As in the analysis of federal reforms, I investigate how sensitive the analysis of state reforms is to model uncertainty. The preceding section showed four different specifications, but focused on one estimation sample (all single women) and extensive margin outcome (weekly employment). This section shows results for all possible permutations of four samples (all single women, low-educated single women, and single women below the median or 25th percentile of predicted earnings) and four outcomes (employment and participation at the weekly or annual levels) in each of the four different specifications (difference-in-differences and triple-differences for all reforms or the largest reforms alone). This gives a total of 64 specifications.

The online appendix presents event studies for all 64 specifications.⁵³ Figure 12 summarizes the findings by showing the distribution of reduced-form estimates (Panel A) and elasticity estimates (Panel B). These are estimates of average three-year effects of state reforms. As can be seen from the figure, the estimates are symmetrically distributed around zero, with the mean being slightly negative. What is more, the event study graphs in appendix reveal that the positive outliers do not in general come from the specifications one would expect: some of the largest estimates are

⁵²This is done separately for single women with children and single women without children. Allowing for separate synthetic control states for the two groups improves pre-trends.

⁵³See Figures A.82-A.97.

based on *all* reforms and the sample of *all* single women, i.e. specifications where the average tax treatment is relatively small. Overall, this analysis suggests that state EITC supplements, even when zooming in on the largest ones, have had no impact on extensive margin labor supply for single mothers.

7 Dissecting the 1990s

The large employment increase among single mothers following the federal EITC expansion in 1993 is a historical anomaly. No other EITC reform at the state or federal level is associated with labor supply changes of such a magnitude. Because of the confounding effects of welfare reform and the business cycle in the mid-late 1990s, it would be natural to conclude that we should ignore the 1993 reform as a means to identifying the impact of the EITC. The reform is not a clean quasi-experiment. Nevertheless, it is interesting to dive deeper into the empirical patterns of the 1990s and attempt to tease apart the EITC from confounding factors. There are two reasons for this. First, a complete story about the EITC and extensive margin labor supply should reconcile the small or zero effects from outside the 1993 reform with the large increases observed after the 1993 reform. Second, because the employment increases for single mothers in the 1990s were so large and historically unique, understanding what happened is important in its own right. In this section, I present a number of analyses intended to show if these employment increases could be consistent with a zero effect of the EITC. The analyses suggest that they could: the employment increases align more closely with welfare reform and business cycle variation than with EITC variation.

7.1 A Needle in a Haystack

The first exercise provides a simple reality check of the data. Here I present simulations showing how much of the employment increases in the mid-late 1990s could be explained by the EITC under different elasticity scenarios. The results will show that trying to detect EITC impacts using the 1993 expansion is akin to looking for a needle in a haystack. The employment changes are so large that, even under sizable tax elasticities, almost all of the variation must be driven by non-tax factors.

From the definition of the elasticity ε in equation (3), the extensive margin response to the 1993 tax reform can be written as

$$\Delta P_t = \varepsilon \cdot \frac{\Delta \left(1 - \tau_t\right)}{1 - \tau_{93}} \cdot P_{93},\tag{4}$$

where ΔP_t is the response in year *t* relative to the pre-reform year 1993, $\Delta (1 - \tau_t)$ is the reforminduced change in the net-of-tax rate in year *t* relative to 1993, while τ_{93} and P_{93} are baseline values in 1993.⁵⁴ Calculating $\Delta (1 - \tau_t)$ as the difference-in-differences between treatment and control groups, the implied time path of ΔP_t can be compared to the observed difference-in-differences impacts from the event studies.

Given the elasticity, the main input into equation (4) is the percentage effect of the reform on the net-of-tax rate, $\frac{\Delta(1-\tau_t)}{1-\tau_{93}}$. In the elasticity estimations presented above, two approximations were made when calculating this entity: the average tax rate did not account for the welfare system, and single women were assumed to enter the labor market at the first kink of the federal EITC. These approximations were made to sidestep data constraints when calculating elasticities across many reforms over a long time period. In the simulation exercise presented here, I take an exact approach that accounts for (i) taxes paid *and* benefits lost in the calculation of net-of-tax rates and (ii) the entire distribution of earnings conditional on working. The tax calculations are based on NBER's tax simulation model (TAXSIM), while the benefit calculations are based on a model of Aid to Families with Dependent Children (AFDC) and Food Stamps (FS). The full details of these calculations are provided in appendix sections C and D.

To calculate tax rates on participation, earnings conditional on working are set equal to observed earnings for workers and predicted earnings for non-workers. The earnings prediction is based on equation (1) run on a pre-reform sample of single women with positive earnings, using the estimated parameters to predict earnings for non-workers. Combining these earnings measures with the tax-benefit simulation model gives the reform-induced change in the net-of-tax rate on participation, $\frac{\Delta(1-\tau_t)}{1-\tau_{93}}$. While the baseline participation tax rate τ_{93} includes *all* taxes on labor income (federal income taxes, federal payroll taxes, and state income taxes) as well as the implicit tax from welfare benefits, the *change* in the participation tax rate $\Delta \tau_t$ accounts only for *federal income taxes* (including but not limited to the EITC). Because the goal is to simulate the impact of the federal tax reform act of 1993, the tax rates from federal payroll taxes, state income taxes, and welfare benefits are held constant at their 1993 levels. The calculation of federal income taxes in each year is based on 1993 earnings (adjusted for inflation) to isolate the mechanical effect of the reform on tax rates.

Figure 13 compares actual and simulated DiD event studies of the 1993 reform. The actual DiD

⁵⁴Note that, while the elasticity formula in (3) is a definition, the expression in (4) is a model: it assumes a static labor supply model in which single women respond according to the constant elasticity ε in each period *t*. The static, iso-elastic model is obviously a simplification, but a natural benchmark to consider given the objective of this exercise.

series plot estimates $\hat{\gamma}_t$ from equation (2) without any controls, while the simulated DiD series plot ΔP_t calculated from equation (4) under an extensive margin elasticity of either 0.25 (short dashes) or 0.5 (long dashes). The smaller of the two elasticities corresponds to the preferred estimate in the meta study by Chetty, Guren, Manoli, and Weber (2013). Results are shown for all single mothers (Panel A) and single mothers with low predicted earnings (Panel B). The main take-away from the figure is that, even if tax elasticities were large, the 1993 tax reform would account for just a small part of the employment increase observed in the mid-late 1990s. Among those most strongly treated by the EITC (Panel B), only 10% (19%) of the employment increase can be explained by the 1993 tax reform under an elasticity of 0.25 (0.50). Hence, most of the variation *must* be driven by confounders, whatever they are. This makes it virtually impossible to identify EITC impacts using the 1993 reform: if the true elasticity were 0.25, we would need to find controls that can absorb 90% of the variation and we can never be confident that we have found the right controls.⁵⁵

7.2 Fanning-Out by Family Size

The difficulty of separating the effects of EITC and welfare reform in the 1990s is that their timing overlapped and that both of them treated single mothers. This difficulty may be addressable, however, because the two policies introduced different treatment variation *within* the group of single mothers. The following sections investigate if such variation can be used to tease apart the two effects. I start by focusing on variation by number of children.

The 1993 EITC expansion was much larger for families with two children than for families with one child, but the same across families with two or more children (see Figures 1-2). Hence, we would expect any EITC-driven divergence in the employment effects across family sizes to occur primarily between one and two children, with little additional divergence between two and three or more children. State and federal welfare reform restricted benefit access across the board, implying that single mothers who were more reliant on welfare benefits were more strongly treated. Because pre-reform welfare participation was sharply increasing in the number of children (as was the post-reform drop in welfare participation), welfare treatment intensity was strictly increasing in family size.⁵⁶ Hence, we expect welfare-driven divergence in the employment effects to occur at all family sizes.

Figure 14 shows event studies of the 1993 reform by number of children (1, 2, 3, and 4+).

⁵⁵Figures A.98-A.99 in the appendix present simulation results by number of children in both the full sample and the low-earnings sample, showing that the identification challenges described here are even stronger in larger families.

⁵⁶See Figure A.3 in the online appendix.

The figure plots DiD estimates $\hat{\gamma}_t^n$ from an extension of equation (2) with separate dummies for each number of children *n*, without any controls in Panel A and with demographic controls in Panel B. In both panels, there is a clear fanning-out of employment effects by number of children. Consider first the raw patterns in Panel A. In the years following the 1993 reform, the employment rate increased by about 10pp for single women with one child, 15pp for single women with two children, 20pp for single women with three children, and almost 30pp for those with four or more children. Because the baseline level of employment is declining in family size, the fanning-out is even more dramatic in percentage terms. For example, single mothers with 4+ children almost doubled their employment rate, from about 30% to about 60%. Panel B of the figure shows that some of the fanning-out can be explained by changes in demographic composition. The fanning-out is still strong, however, with increases in the employment rate ranging from about 8pp to 25pp across family sizes. Appendix Figures A.102-A.103 reproduce the analysis for the other extensive margin measures, showing that the fanning-out by family size is a robust feature of the data.

These patterns of heterogeneity are consistent with a welfare reform interpretation, but harder to explain with an EITC interpretation. It is possible that extensive margin elasticities are increasing in family size, which would generate a fanning-out of employment responses even to the EITC. However, heterogeneity in elasticities is not a plausible explanation for the empirical patterns documented here simply because the amount of heterogeneity is too large. An earlier version of this paper (Kleven 2020) calculated elasticities by family size to demonstrate this point. According to these calculations, if the fanning-out is attributed solely to the EITC, the extensive margin elasticity would have to be 1.4 for single mothers with two children, 2.3 for single mothers with three children, and 4.1 for single mothers with four or more children.⁵⁷ Such elasticity heterogeneity is implausible.

7.3 Impacts by Welfare Treatment Intensity

The analysis of heterogeneity by number of children is based on the idea of using variation in welfare treatment intensity (relative to EITC treatment intensity) within the group of single mothers to tease apart the impacts of welfare and EITC reform. This section pursues this idea in greater depth, considering two alternative proxies for welfare treatment intensity. The first proxy is the age of the

⁵⁷Similar magnitudes can be inferred from the simulation analysis by family size in Appendix Figure A.98. This figure shows, for each family size, what fraction of the employment increase in the 1990s can be explained by the EITC under specific elasticities. These results can be used to back out what the elasticity would have to be for the EITC to explain 100% of the employment increase. This gives roughly similar (somewhat larger) elasticities than those reported above. The estimates are not exactly the same in part because Figure A.98 is based on raw data, without any controls.

youngest child. It turns out that the level of welfare participation prior to waivers and PRWORA and the ensuing drop in welfare participation are strongly related to the age of the youngest child. The second proxy is a predicted probability of pre-reform welfare participation using the age of the youngest child, the number of children, and other demographic variables. Both of these analyses will show that the extensive margin effects in the 1990s are closely aligned with the strength of welfare treatment.

Due to the granularity of the analysis in this section, rather than showing event study graphs, I summarize the effects using a DiD specification with a post-reform dummy. Specifically, I consider specifications of the following form

$$P_{imt} = \alpha \cdot Post_t + \sum_{j} \beta_j \cdot Welfare_{j=c} + \sum_{j} \gamma_j \cdot Post_t \cdot Welfare_{j=c} + \mathbf{X}_i \boldsymbol{\phi} + \eta \cdot U_{st} + \theta \cdot U_{st} \cdot Kids_i + \nu_{imt},$$
(5)

where P_{imt} is an indicator equal to one if individual *i* is employed in month *m* of year *t*, $Post_t$ is an indicator equal to one in the years after the 1993 reform, and $Welfare_{j=c}$ is an indicator equal to one if the individual belongs to welfare treatment category *c*. The welfare treatment categories are based on either the age of the youngest child (7 bins) or a predicted AFDC probability (10 deciles). In either case, the omitted category is having no children such that the welfare category variable subsumes the kids dummy from the previous specifications. The coefficient γ_c represents the average DiD effect for single mothers in welfare category *c* relative to single women without children over a specified time horizon. I consider two different time horizons: a three-year horizon (avoiding years after PRWORA) and a ten-year horizon. I start from raw DiD effects obtained from the first line of equation (5), and then consider the implications of adding the controls in the second line. These controls absorb the effects of demographics and state-level business cycles, specified as in equation (2) above.

The results are presented in Figure 15. The left panels show heterogeneity by age of the youngest child, while the right panels show heterogeneity by probability of AFDC participation. Consider first the results for age of the youngest child.⁵⁸ The top panel shows that this variable is a strong predictor of welfare participation before the reform — the younger is the child, the higher is participation — which in turn predicts the drop in welfare participation after the reform. The

⁵⁸As elsewhere, I include only EITC-eligible children and consider the following categories for the age of the youngest child: 0-1, 2-3, 4-6, 7-9, 10-13, 14-17, and 18+. Consistent with the EITC rules, the 18+ category includes children aged 18-23, who are living at home and are still in full-time education.

relationship between pre-reform participation levels and post-reform participation drops across bins of the age of the youngest child is almost perfectly linear and has a slope of 0.69. Among single mothers with children aged 0-1, pre-reform AFDC participation was about 50 percent and the subsequent fall was close to 40 percentage points. By contrast, single mothers with older children had very low AFDC participation rates and, as a result, were virtually unaffected by welfare reform.

The middle panel shows raw DiD estimates of the average employment effect over three years (solid black) and ten years (dashed black) by age of the youngest child. It also shows the average drop in the welfare participation rate over ten years (dashed red), estimated from equation (5) using an indicator for welfare receipt as the outcome variable. The extensive margin effect is monotonically declining in the age of the youngest child. The three-year effect falls to about zero (and becomes statistically insignificant) for those whose youngest child is older than 13 years, while the ten-year effect falls to zero when the youngest child is older than 17. These estimates include any confounding effects of changing demographics and the business cycle. As shown in the bottom panel, controlling for these factors reduces the treatment effect at each age and therefore strengthens the results. The three-year effect disappears above age 6 and the ten-year effect disappears above age 13.

Because the analysis is based on EITC-eligible children — children living at home who are below age 18 or below age 24 if still in education — the age of the youngest child is correlated with the number of children. Having more children imply that the last-born tends to be younger. As a result, the analysis using age of youngest child overlaps with the previous analysis using number of children. I now move to a multivariate analysis using the joint predictive power of the age of youngest child, the number of children, and other demographic variables. Specifically, I estimate the probability of pre-reform welfare participation based on the following specification

$$AFDC_i = \alpha_a + \beta_n + \gamma_y + \zeta_r + \lambda_s + \nu_i, \tag{6}$$

where an indicator for receiving welfare benefits, $AFDC_i$, is regressed on fixed effects for the age of the mother *a*, the number of children *n*, the age of the youngest child *y*, race *r*, and state *s*.⁵⁹ The

⁵⁹Equation (6) does not include education even though this variable would help with predicting AFDC participation. Education is excluded because it is a relatively direct proxy for earnings conditional on working and therefore for EITC eligibility. As a result, while using education information would strengthen my results, this is misleading as highly-educated single mothers who did not participate in welfare (thus being untreated by welfare reform) were also less likely to be eligible for the EITC (thus being untreated by EITC reform as well).

binning of these demographic variables is the same as before. The regression is run on the CPS March files from 1994, which contains information on welfare receipt in the pre-reform year, 1993. From this regression, I predict the probability of AFDC participation for each single mother in the estimation sample, and define an indicator for being in different deciles of the distribution of these AFDC probabilities.

The results are presented in the right panels of Figure 15. The top panel shows that prereform welfare participation varies widely across the distribution, from 8% to more than 60%, and is strongly related to the post-reform welfare participation drop. The relationship between pre-reform levels and post-reform drops is stronger in this specification than when using only the age of the youngest child. The next panels plot the average employment effect over three years and ten years by decile of predicted AFDC probability. The employment effect is strongly increasing in the AFDC probability. In the raw data, the ten-year effect is almost 20pp in the top decile, but only 2pp in the bottom decile. Adding controls for demographics and state-level business cycles makes the estimates smaller: the ten-year effect is about 12pp in the top decile and negative in the bottom decile. In fact, there are no significant ten-year effects in the bottom two deciles and no significant three-year effects in the bottom four deciles.⁶⁰

To summarize, the large extensive margin increases in the 1990s were driven by single mothers who had high AFDC participation prior to welfare reform and were therefore strongly treated by waivers and PRWORA. Single mothers with low pre-reform AFDC participation did not respond. These results are consistent with welfare reform, but harder to reconcile with EITC reform.

7.4 The Impact of Welfare Waivers

The confounding effects of welfare reform operate at both the state level (waivers) and the national level (PRWORA). The preceding sections used heterogeneity in welfare treatment intensity in an attempt to control for both, but such patterns of heterogeneity are subject to endogeneity concerns. This section asks how much of the employment increase in the 1990s can be explained by welfare waivers using a quasi-experimental approach. Waivers were introduced in some states but not others, creating plausibly exogenous variation in welfare benefits. The main focus is on the years 1994-96 — after the 1993 EITC reform, but before nationwide TANF reform — where we have the best chance of separating the effects of welfare and EITC reform.

⁶⁰Appendix Figures A.104-A.107 reproduce the analysis for all four extensive margin outcomes, showing that the results are robust.

To investigate the confounding effect of welfare waivers, I conduct event studies of the 1993 EITC reform separately for states that approved statewide waiver legislation and for those that did not. As shown in Table A.3 of the appendix, there were 13 states without any statewide waiver legislation: Alabama, Alaska, District of Columbia, Kansas, Kentucky, Louisiana, Nevada, New Mexico, New York, Oklahoma, Pennsylvania, Rhode Island, and Wyoming. Figure 16 shows event studies for all non-waiver states (blue series) and all waiver states (black series). These series plot DiD coefficients from specification (2), implemented separately on the two groups of states, controlling for demographics and state-level business cycles.

The figure shows that the employment effect is much smaller in non-waiver states than in waiver states in the years before nationwide TANF reform. In fact, the non-waiver effect is close to zero and statistically insignificant between 1994-1996. Following the implementation of TANF reform, non-waiver states begin to catch up to waiver states and converge to the same long-run effect. As a result, restricting attention to non-waiver states changes the DiD series from what looks like an event study of the 1993 EITC reform to what looks like an event study of the 1996 TANF reform. These patterns are exactly consistent with welfare reform being the main driver of the employment increases of single mothers in the 1990s. Figure A.108 replicates the analysis for different extensive margin outcomes (weekly vs annual employment) and for different estimation samples (all single women, low-educated single women, and single women with low predicted earnings). There is some variation across the different specifications, but the fundamental qualitative insight is robust. If anything, the results are even stronger when considering annual employment, and they survive in the low-educated and low-earnings samples.

7.5 Changing Social Norms

Another possibility is that social norms regarding welfare receipt changed in the 1990s. The public debate certainly changed, featuring increasingly derogatory rhetoric and racial stereotypes to describe welfare mothers.⁶¹ As society became less tolerant of welfare mothers, their willingness to work may have increased above and beyond any effect of economic incentives. Conceptually, there are two effects of social norms to consider: (i) Norm changes may have precipitated welfare reform (and EITC reform) and at the same time changed behavior, leading researchers to attribute an impact of norms to policies; (ii) welfare reform may have been perceived as a social injunction

⁶¹As an example of the tenor of the debate, Figure A.109 shows an (infamous) cover of *The New Republic* on August 12, 1996, just before PRWORA was signed into law by President Bill Clinton.

that changed norms, creating a social multiplier effect of the reform. In the latter case, there would be no bias in the reduced-form impact of the reform (leaving aside other confounders), but there would be bias in the inferred elasticity with respect to the net-of-tax rate.

Figure A.110 in the appendix suggests that social norms may have been a confounding factor in the 1990s. Using Google Ngrams, the figure traces the use of loaded terms to describe welfare recipients in books published in the United States since 1970. Panel A considers the term *welfare queen*, a derogatory phrase used to describe women on welfare.⁶² Panel B considers the terms *undeserving poor* and *deserving poor*. The graphs demonstrate that the use of negative language surrounding welfare receipt and poverty exploded in the 1990s, consistent with important changes to social norms and culture. Therefore, trying to understand the historic labor supply shift for single mothers in the 1990s solely through the lens of economic incentives may be too narrow.

8 Conclusion

The EITC program is widely credited for having had large effects on extensive margin labor supply, especially for single mothers. This paper reappraises the difference-in-differences and event study approaches that underpin much of this consensus. The paper investigates every EITC reform at the state and federal level since the inception of the policy in 1975. All reforms are analyzed in an event study framework and a comprehensive analysis of model uncertainty is conducted. The analysis suggests that the EITC has not had sizable and robust effects on extensive margin labor supply. Apart from the federal 1993 reform, most specifications imply small and statistically insignificant effects. The 1993 reform, on the other hand, is associated with large effects, but the interpretation is complicated due to contemporaneous changes in the welfare system and the macroeconomy. Different analyses are presented to suggest that the EITC. For example, controlling for the business cycle, non-waiver states did not see any significant employment effects in the years immediately after federal EITC reform, before the implementation of federal welfare reform.

The main take-away of the paper is that quasi-experimental analyses of EITC reform are fragile. Large EITC elasticities require researchers to restrict attention to particular reforms (some of which are heavily confounded) and particular econometric specifications, ignoring many other reforms and specifications that are no less valid. While the paper demonstrates the fragility of prior esti-

⁶²The phrase was coined in the Chicago Tribune in 1974 and later popularized by Ronald Reagan.

mates, it does not conclusively prove that the true response is zero. There are two reasons for this. First, event study and difference-in-differences designs capture only short-run effects of the EITC. It is possible that short-run effects are attenuated due to optimization frictions, including informational and psychological frictions. A number of studies have documented that most potential EITC recipients have limited awareness and understanding of the program, especially around the historical expansions studied in the literature.⁶³ At the same time, although optimization frictions seem important in this context, the long-run descriptive evidence presented in the paper is not *prima facie* suggestive of gradual long-run responses.

Second, empirical designs that define treatment status based on fertility rely on strong assumptions and raise concerns about confounding shocks. A cautious interpretation of my findings is to view them as a demonstration that we learn relatively little from the quasi-experimental EITC literature due to limitations in the empirical designs. This calls for focusing on randomized trials, especially those that consider policies similar to the EITC. The closest we have to such trials are the Paycheck Plus demonstrations in New York City and Atlanta (Miller, Katz, Azurdia, Isen, Schultz, and Aloisi 2018; Yang, Bernardi, Metz, Miller, Katz, and Isen 2022). These trials found either modest or zero effects on extensive margin labor supply, consistent with the conclusions in this paper. While such experiments resolve concerns about causal identification, they raise external-validity concerns and also capture only short-run effects. Attenuation bias from information frictions is less obvious, however, as information and personal outreach were provided as part of the experiments.

To interpret the starkly different effects of EITC and welfare reform estimated here, it is worth noting that welfare reform is less vulnerable to concerns about salience and information. The debate about welfare reform and welfare culture was extremely prominent in the nineties, as exemplified by Bill Clinton's famous campaign pledge to "end welfare as we know it." State waivers and national TANF reform imposed drastic changes on *existing* program participants, so information was essentially automatic. Moreover, the changes to welfare were largely related to ordeals and enforcement, the impact of which is more mechanical than financial incentives. They included lifetime limits on welfare receipt, work requirements, community service and training, and those who did not satisfy these requirements could get kicked off the welfare rolls. These initiatives stand in sharp contrast to the complex tax refund incentives introduced by the relatively unknown EITC program.

⁶³See e.g., Romich and Weisner (2000), Smeeding, Phillips, and O'Connor (2000), Phillips (2001), Maag (2005), Chetty and Saez (2013), and Bhargava and Manoli (2015).

Finally, the findings in this paper — even if we interpret them as a conclusive zero effect — do not necessarily imply that the EITC is a bad policy. Absent labor supply responses at either the extensive or intensive margins, the EITC is a non-distortionary transfer to the working poor.⁶⁴ The optimality of such a transfer depends on the social welfare function and on who pays for it. For example, if the EITC is financed by a lump-sum tax on all individuals and if the government puts a larger weight on the working poor than on the average individual, then the policy is socially optimal. This is the reasoning underlying the results in Saez (2002).⁶⁵ Conversely, if transfers to the working poor are financed by reducing welfare benefits to the poor, then the equity effect is negative under standard social preferences. In this case, the desirability of the EITC requires social preferences that put more weight on the working poor than on the non-working poor, even if the latter are worse off. Leaving aside these optimal tax considerations, the political argument for the EITC has relied heavily on the idea of large employment effects. This argument is empirically fragile.

⁶⁴This paper does not investigate intensive margin responses to the EITC. In general, the literature has struggled to find any clear evidence of intensive margin responses to the program, perhaps because of the frictions discussed above. Chetty, Friedman, and Saez (2013) is an exception in the literature, arguing that EITC eligibles do respond along the intensive margin once information frictions are accounted for.

⁶⁵In Saez (2002), the optimal EITC (defined as a negative tax rate on participation) reflects a trade-off between the equity gain just described and the efficiency *loss* from extensive margin responses to an EITC that make people work too much.

References

- BASTIAN, JACOB (2020): "The Rise of Working Mothers and the 1975 Earned Income Tax Credit," *American Economic Journal: Economic Policy*, 12(3), 44–75. 7, 22, 23
- BERUBE, ALAN, ANNE KIM, BENJAMIN FORMAN, AND MEGAN BURNS (2002): "The Price of Paying Taxes: How Tax Preparation and Refund Loans Erode the Benefits of the EITC," Brookings Institution and the Progressive Policy Institute. 9
- BHARGAVA, SAURABH, AND DAYANAND MANOLI (2015): "Psychological Frictions and the Incomplete Take-Up of Social Benefits: Evidence from an IRS Field Experiment," *American Economic Review*, 105(11), 1–42. 6, 9, 38
- BLANK, REBECCA M. (2002): "Evaluating Welfare Reform in the United States," *Journal of Economic Literature*, 40(4), 1105–1166. 3, 7, 8
- BOLLINGER, CHRISTOPHER, BARRY HIRSCH, CHARLES HOKAYEM, AND JAMES ZILIAK (2019): "Trouble in the Tails? What We Know about Earnings Nonresponse Thirty Years after Lillard, Smith, and Welch," *Journal of Political Economy*, 127(5), 2143–2185. 14
- CARD, DAVID, AND DEAN R. HYSLOP (2005): "Estimating the Effects of a Time-Limited Earnings Subsidy for Welfare-Leavers," *Econometrica*, 73(6), 1723–1770. 5, 8
- CHETTY, RAJ (2012): "Bounds on Elasticities With Optimization Frictions: A Synthesis of Micro and Macro Evidence on Labor Supply," *Econometrica*, 80(3), 969–1018. 3
- CHETTY, RAJ, JOHN N. FRIEDMAN, TORE OLSEN, AND LUIGI PISTAFERRI (2011): "Adjustment Costs, Firm Responses, and Micro vs. Macro Labor Supply Elasticities: Evidence from Danish Tax Records," *The Quarterly Journal of Economics*, 126(2), 749–804. 3
- CHETTY, RAJ, JOHN N. FRIEDMAN, AND EMMANUEL SAEZ (2013): "Using Differences in Knowledge across Neighborhoods to Uncover the Impacts of the EITC on Earnings," *American Economic Review*, 103(7), 2683–2721. 8, 9, 14, 39
- CHETTY, RAJ, ADAM GUREN, DAY MANOLI, AND ANDREA WEBER (2013): "Does Indivisible Labor Explain the Difference between Micro and Macro Elasticities? A Meta-Analysis of Extensive Margin Elasticities," *NBER Macroeconomics Annual*, 27(1), 1–56. 5, 8, 26, 31
- CHETTY, RAJ, AND EMMANUEL SAEZ (2013): "Teaching the Tax Code: Earnings Responses to an Experiment with EITC Recipients," *American Economic Journal: Applied Economics*, 5(1), 1–31. 9, 38
- COUNCIL OF ECONOMIC ADVISORS (1997): "Explaining the Decline in Welfare Receipt, 1993-1996," Washington: Executive Office of the President. Retrived at https:// clintonwhitehouse4.archives.gov/WH/EOP/CEA/Welfare/Report.html. 12
 - (1999): "The Effects of Welfare Policy and the Economic Expansion on Welfare

Caseloads: An Update," Washington: Executive Office of the President. Retrived at https: //clintonwhitehouse4.archives.gov/WH/EOP/CEA/html/welfare/. 12

- DEPARTMENT OF HEALTH AND HUMAN SERVICES (1999): "State Implementation of Major Changes to Welfare Policies, 1992-1998," Washington, DC: US Department of Health and Human Services. Assistant Secretary for Planning and Evaluation. Retrieved at https://aspe.hhs. gov/report/state-implementation-major-changes-welfare-policies-1992-1998. 12, 71
- EISSA, NADA, AND HILARY HOYNES (2004): "Taxes and the Labor Market Participation of Married Couples: The Earned Income Tax Credit," *Journal of Public Economics*, 88(9–10), 1931–1958. 22
- (2006): "Behavioral Responses to Taxes: Lessons from the EITC and Labor Supply," *Tax Policy and the Economy*, 20, 73–110. 7
- EISSA, NADA, HENRIK J. KLEVEN, AND CLAUS T. KREINER (2006): "Welfare Effects of Tax Reform and Labor Supply at the Intensive and Extensive Margins," in *Tax Policy and Labor Market Performance*, ed. by Jonas Agell, and Peter Sorensen. MIT Press: Cambridge, MA. 10, 20
- (2008): "Evaluation of Four Tax Reforms in the United States: Labor Supply and Welfare Effects for Single Mothers," *Journal of Public Economics*, 92(3–4), 795–816. 10, 20
- EISSA, NADA, AND JEFFREY B. LIEBMAN (1993): "The End of Welfare as We Know It? Behavioral Responses to the Earned Income Tax Credit," *Department of Economics, Harvard University, mimeo*.
 9
- (1996): "Labor Supply Response to the Earned Income Tax Credit," *Quarterly Journal of Economics*, 111(2), 605–637. 1, 7, 9, 26, 27, 57, 72
- ELLWOOD, DAVID T. (2000): "The Impact of the Earned Income Tax Credit and Social Policy Reforms on Work, Marriage, and Living Arrangements," *National Tax Journal*, 53(4), 1063–1105.
 3
- FANG, HANMING, AND MICHAEL P. KEANE (2004): "Assessing the Impact of Welfare Reform on Single Mothers," *Brookings Papers on Economic Activity*, 2004(1), 1–116. 3, 9
- GELBER, ALEXANDER, AND JOSHUA MITCHELL (2011): "Taxes and Time Allocation: Evidence from Single Women and Men," *The Review of Economic Studies*, 79(3), 863–897. 7
- GROGGER, JEFFREY (2003): "The Effects of Time Limits, the EITC, and Other Policy Changes on Welfare Use, Work, and Income Among Female-Headed Families," *Review of Economics and Statistics*, 85(2), 394–408. 3, 9
- GROGGER, JEFFREY, AND LYNN A. KAROLY (2005): *Welfare Reform: Effects of a Decade of Change*. Harvard University Press: Cambridge, MA. 5, 7, 8, 12
- HOTZ, V. JOSEPH, CHARLES H. MULLIN, AND JOHN KARL SCHOLZ (2006): "Examining the Effect

of the Earned Income Tax Credit on the Labor Market Participation of Families on Welfare," NBER Working Paper No. 11968. 7

- HOTZ, V. JOSEPH, AND JOHN KARL SCHOLZ (2003): "The Earned Income Tax Credit," in *Means-Tested Transfer Programs in the United States*, ed. by Robert A. Moffitt. Chicago: University of Chicago Press. 7
- HOYNES, HILARY, AND ANKUR PATEL (2018): "Effective Policy for Reducing Poverty and Inequality? The Earned Income Tax Credit and the Distribution of Income," *Journal of Human Resources*, 53(4), 859–890. 1, 7
- JONES, DAMON (2010): "Information, Preferences and Public Benefit Participation: Experimental Evidence from the Advance EITC and 401(k) Savings," *American Economic Journal: Applied Economics*, 2(2), 147–63. 9
- KLEVEN, HENRIK J. (2016): "Bunching," Annual Review of Economics, 8, 435–464. 3
- (2020): "The EITC and the Extensive Margin: A Reappraisal," Working Paper, February 2020. 32
- (2021): "Sufficient Statistics Revisited," Annual Review of Economics, 13, 515–538. 26
- (2023): "The Geography of Child Penalties and Gender Norms: A Pseudo-Event Study Approach," NBER Working Paper No. 30176. 2, 8
- KLEVEN, HENRIK J., AND CLAUS T. KREINER (2005): "Labor Supply Behavior and the Design of Tax and Transfer Policy," *Danish Journal of Economics*, 143, 321–358. 20
- (2006): "The Marginal Cost of Public Funds: Hours of Work Versus Labor Force Participation," *Journal of Public Economics*, 90(10–11), 1955–1973. 10
- KLEVEN, HENRIK J., CLAUS T. KREINER, KRISTIAN LARSEN, AND JAKOB E. SØGAARD (2023): "Micro vs Macro Labor Supply Elasticities: The Role of Dynamic Returns to Effort," NBER Working Paper No. 31549. 4
- KLEVEN, HENRIK J., CAMILLE LANDAIS, JOHANNA POSCH, ANDREAS STEINHAUER, AND JOSEF ZWEIMÜLLER (2019): "Child Penalties Across Countries: Evidence and Explanations," *AEA Papers and Proceedings*, 109, 122–126. 8

(2022): "Do Family Policies Reduce Gender Inequality? Evidence from 60 Years of Policy Experimentation," *American Economic Journal: Economic Policy*, Forthcoming. 8

- KLEVEN, HENRIK J., CAMILLE LANDAIS, AND JAKOB E. SØGAARD (2019): "Children and Gender Inequality: Evidence from Denmark," *American Economic Journal: Applied Economics*, 11(4), 181– 209. 2, 8
- KLINE, PATRICK, AND MELISSA TARTARI (2016): "Bounding the Labor Supply Responses to a

Randomized Welfare Experiment: A Revealed Preference Approach," *American Economic Review*, 106(4), 972–1014. 5, 9

- KOPCZUK, WOJCIECH, AND CRISTIAN POP-ELECHES (2007): "Electronic Filing, Tax Preparers and Participation in the Earned Income Tax Credit," *Journal of Public Economics*, 91(7–8), 1351–1367. 9
- MAAG, ELAINE (2005): "Paying the Price? Low-Income Parents and the Use of Paid Tax Preparers.," New Federalism: National Survey of America's Families, Series B, No. 64. 9, 38
- MARTINEZ, ISABEL, EMMANUEL SAEZ, AND MICHAEL SIEGENTHALER (2021): "Intertemporal Labor Supply Substitution? Evidence from the Swiss Income Tax Holidays," *American Economic Review*, 111(2), 506–546. 5
- MEAD, LAWRENCE (2014): "Overselling the Earned Income Tax Credit," *National Affairs*, 21, 20–33. 9
- MEYER, BRUCE (2010): "The Effects of the Earned Income Tax Credit and Recent Reforms," in *Tax Policy and the Economy, Volume 24*, ed. by Jeffrey R. Brown. Chicago: University of Chicago Press. 1, 7
- MEYER, BRUCE, AND DAN ROSENBAUM (2001): "Welfare, The Earned Income Tax Credit, and the Labor Supply of Single Mothers," *Quarterly Journal of Economics*, 116(3), 1063–1114. 1, 3, 7, 9
- MILLER, CYNTHIA, LAWRENCE F. KATZ, GILDA AZURDIA, ADAM ISEN, CAROLINE B. SCHULTZ, AND KALI ALOISI (2018): "Boosting the Earned Income Tax Credit for Singles: Final Impact Findings From the Paycheck Plus Demonstration in New York City," *New York: MDRC.* 6, 8, 38
- MURRAY, CHARLES (1984): Losing Ground: American Social Policy, 1950-1980. Basic Books: New York, NY. 1
- NICHOLS, AUSTIN, AND JESSE ROTHSTEIN (2015): "The Earned Income Tax Credit," in *The Economics of Means-Tested Transfer Programs in the United States, Volume 1*, ed. by Robert A. Moffitt. Chicago: University of Chicago Press. 7
- OLSON, LYNN M., AND AUDREY DAVIS (1994): "The Earned Income Tax Credit: Views from the Street Level," *Center for Urban Affairs and Policy Research*. 9
- PHILLIPS, KATHERIN ROSS (2001): "Who Knows About the Earned Income Tax Credit?," New Federalism: National Survey of America's Families, Series B, No. 64. 9, 38
- ROMICH, JENNIFER L., AND THOMAS WEISNER (2000): "How Families View and Use the EITC: Advance Payment versus Lump Sum Delivery," *National Tax Journal*, 53(4), 1245–1265. 9, 38
- SAEZ, EMMANUEL (2002): "Optimal Income Transfer Programs: Intensive Versus Extensive Labor Supply Responses," *Quarterly Journal of Economics*, 117(3), 1039–1073. 20, 39
- SAEZ, EMMANUEL, MANOS MATSAGANIS, AND PANOS TSAKLOGLOU (2012): "Earnings Determi-

nation and Taxes: Evidence From a Cohort-Based Payroll Tax Reform in Greece," *The Quarterly Journal of Economics*, 127(1), 493–533. 5

- SAEZ, EMMANUEL, JOEL B. SLEMROD, AND SETH H. GIERTZ (2012): "The Elasticity of Taxable Income with Respect to Marginal Tax Rates: A Critical Review," *Journal of Economic Literature*, 50(1), 3–50. 5
- SCHANZENBACH, DIANE WHITMORE, AND MICHAEL R. STRAIN (2021): "Employment Effects of the Earned Income Tax Credit: Taking the Long View," *Tax Policy and the Economy*, 35(1), 87–129.
 7, 26, 57, 72
- SMEEDING, TIMOTHY, KATHERIN ROSS PHILLIPS, AND MICHAEL O'CONNOR (2000): "The EITC: Expectation, Knowledge, Use, and Economic and Social Mobility," *National Tax Journal*, 53(4), 1187–1209. 9, 38
- YAGAN, DANNY (2019): "Employment Hysteresis from the Great Recession," *Journal of Political Economy*, 127(5), 2505–2558. 17
- YANG, EDITH, ALEXANDRA BERNARDI, RACHAEL METZ, CYNTHIA MILLER, LAWRENCE F. KATZ, AND ADAM ISEN (2022): "An Earned Income Tax Credit That Works for Singles: Final Impact Findings From the Paycheck Plus Demonstration in Atlanta," *New York: MDRC.* 6, 8, 38

	Single	e Women Wi	th Children	Single Women Without Children				
	All	Low Education	Low Predicted Earnings	All	Low Education	Low Predicted Earnings		
	(1)	(2)	(3)	(4)	(5)	(6)		
Weekly Employment Rate	0.68	0.60	0.53	0.75	0.66	0.68		
Annual Employment Rate	0.73	0.66	0.63	0.81	0.74	0.78		
Earnings	22,186	15,313	10,923	25,924	18,545	13,082		
Age	34.37	33.26	25.51	30.67	32.15	21.55		
Fraction White	0.68	0.67	0.58	0.79	0.75	0.79		
Fraction Black	0.26	0.27	0.34	0.14	0.18	0.14		
Number of Children	1.79	1.89	2.12	0.00	0.00	0.00		
Age of Youngest Child	7.51	7.11	3.11					
High School & Below	0.55	1.00	0.79	0.33	1.00	0.42		
College Degree	0.13	0.00	0.00	0.31	0.00	0.00		
Observations	1,787,348	979,702	447,099	3,010,586	993,063	765,901		

TABLE 1: DESCRIPTIVE STATISTICS

Notes: This table shows means of demographic and labor market variables among single women with and without EITC-eligible children, respectively, where EITCeligible children are those under the age of 19 (24 if in full-time education) and living at home for at least half of the year. The columns consider different samples: all single women, single women with low education (high school degree or below), and single women with low predicted earnings (bottom quartile of the within-year distribution of predicted earnings, estimated using equation 1). All samples include single women aged 20-50 using the March and monthly CPS files combined, pooling all years from 1968 to 2019. Earnings are reported in 2018 USD.

	Weekly	Employm	ent Rate	Annual	Employm	ent Rate
	(1)	(2)	(3)	(4)	(5)	(6)
	Pane	l A: 1975 R	eform			
Effect of Reform (pp)	-0.71	-0.83	0.33	-0.42	-0.66	1.89
	(1.38)	(1.33)	(1.61)	(1.23)	(1.19)	(1.45)
Elasticity	-0.10	-0.12	0.05	-0.05	-0.08	0.24
	(0.19)	(0.19)	(0.23)	(0.15)	(0.15)	(0.18)
	Pane	l B: 1986 R	eform			
Effect of Reform (pp)	-1.03	-0.11	-0.70	1.50	1.77*	1.09
	(1.07)	(1.01)	(1.03)	(0.98)	(0.94)	(0.96)
Elasticity	-0.48	-0.05	-0.33	0.63	0.75*	0.46
	(0.50)	(0.47)	(0.48)	(0.41)	(0.39)	(0.40)
	Pane	l C: 1990 R	eform			
Effect of Reform (pp)	-0.35	-1.73*	-0.39	0.00	-2.53***	-1.84*
	(1.07)	(1.00)	(1.04)	(0.98)	(0.94)	(0.98)
Elasticity	-0.13	-0.65*	-0.15	0.00	-0.86***	-0.62*
-	(0.40)	(0.37)	(0.39)	(0.33)	(0.32)	(0.33)
	Pane	l D: 1993 R	eform			
Effect of Reform (pp)	3.06***	2.81***	1.29***	4.39***	3.72***	2.66***
	(0.51)	(0.47)	(0.49)	(0.98)	(0.92)	(0.96)
Elasticity	0.53***	0.49***	0.22***	0.67***	0.57***	0.41***
	(0.09)	(0.08)	(0.08)	(0.15)	(0.14)	(0.15)
	Pane	1 E: 2009 R	eform			
Effect of Reform (pp)	-3.55***	-4.02***	-3.01***	-1.25	-2.02	-0.58
	(1.07)	(1.03)	(1.14)	(1.70)	(1.67)	(1.86)
Elasticity	-0.70***	-0.79***	-0.59***	-0.23	-0.37	-0.10
	(0.21)	(0.20)	(0.22)	(0.31)	(0.30)	(0.34)
Panel F:	Federal R	eforms Sta	cked, Wit	hout 1993		
Effect of Reform (pp)	-1.41**	-1.67***	-0.94	-0.04	-0.86	0.14
	(0.57)	(0.54)	(0.62)	(0.62)	(0.61)	(0.69)
Elasticity	-0.33**	-0.39***	-0.22	-0.01	-0.18	0.03
	(0.13)	(0.13)	(0.15)	(0.13)	(0.13)	(0.15)
Demographic Controls:		Х	Х		Х	Х
Unemployment Controls:		~	X			X

TABLE 2: DID EFFECTS OF FEDERAL EITC REFORMS ON SINGLE MOTHERS ALL SINGLE WOMEN

Notes: This table reports DiD estimates of the effects of the five federal EITC reforms separately (Panels A-E) and stacked together (Panel F). These are three-year effects, estimated from a modified version of equation (2) in which the first three year dummies after each reform have been collapsed into a post dummy. Panels A-D are based on comparing single women with and without children, while Panel E is based on comparing single women with 3+ children to those without children. The columns show results for different outcomes (weekly and annual employment) and for different controls. The rows show the reduced-form effect of each reform and the implied elasticity with respect to $1 - \tau$ as defined in equation (3). The sample includes all single women aged 20-50. Panels A-C use March CPS files alone, while Panels D-E use March and monthly CPS files combined. Robust standard errors are clustered at the individual level.

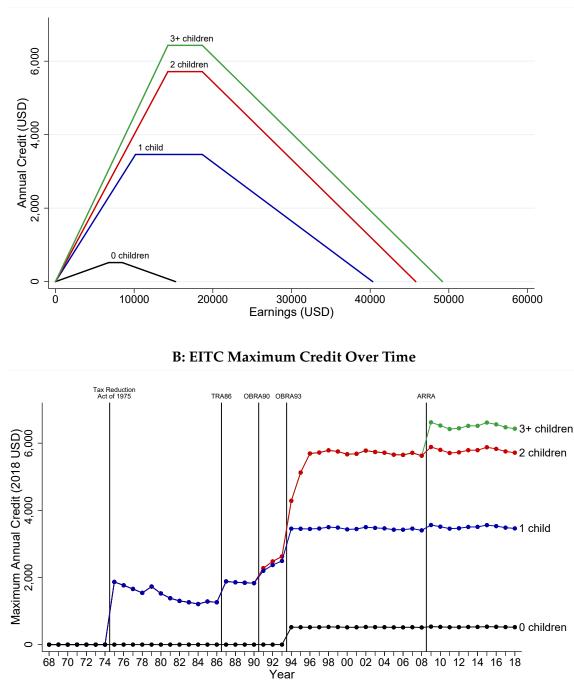
	Weekly	Employm	ent Rate	Annual	Employm	ent Rate
	(1)	(2)	(3)	(4)	(5)	(6)
	Pane	l A: 1975 F	Reform			
Effect of Reform (pp)	1.12	1.01	0.05	0.06	-0.62	3.43
	(2.00)	(1.94)	(2.36)	(1.78)	(1.74)	(2.14)
Elasticity	0.19	0.17	0.01	0.01	-0.09	0.47
	(0.34)	(0.33)	(0.40)	(0.25)	(0.24)	(0.30)
	Pane	l B: 1986 F	leform			
Effect of Reform (pp)	-0.42	0.14	-0.86	3.14**	3.24**	2.23
	(1.59)	(1.53)	(1.57)	(1.45)	(1.40)	(1.44)
Elasticity	-0.26	0.09	-0.53	1.57**	1.61**	1.11
·	(0.98)	(0.94)	(0.97)	(0.72)	(0.70)	(0.72)
	Pane	l C: 1990 F	leform			
Effect of Reform (pp)	-0.79	-1.84	-0.23	-2.64*	-4.18***	-3.62*
·11/	(1.60)	(1.53)	(1.59)	(1.48)	(1.43)	(1.48)
Elasticity	-0.36	-0.84	-0.11	-0.97*	-1.54***	-1.34*
ý	(0.73)	(0.70)	(0.73)	(0.55)	(0.53)	(0.55)
	Pane	l D: 1993 F	Reform			
Effect of Reform (pp)	4.17***	3.67***	1.72**	7.43***	6.69***	5.42**
	(0.75)	(0.71)	(0.74)	(1.51)	(1.43)	(1.50)
Elasticity	0.78***	0.69***	0.32**	1.12***	1.00***	0.81**
	(0.14)	(0.13)	(0.14)	(0.23)	(0.21)	(0.23)
	Pane	1 E: 2009 R	eform			
Effect of Reform (pp)	-2.98**	-3.32**	-2.65*	-0.44	-1.21	0.42
·* * ·	(1.34)	(1.30)	(1.48)	(2.18)	(2.14)	(2.48)
Elasticity	-0.63**	-0.71**	-0.57*	-0.08	-0.23	0.08
·	(0.28)	(0.28)	(0.32)	(0.41)	(0.41)	(0.47)
Panel F: I	Federal R	eforms Sta	acked, Wit	thout 1993	3	
Effect of Reform (pp)	-0.77	-1.00	-0.92	0.03	-0.69	0.62
	(0.82)	(0.79)	(0.90)	(0.87)	(0.85)	(0.98)
Elasticity	-0.22	-0.28	-0.26	0.01	-0.16	0.14
	(0.23)	(0.22)	(0.25)	(0.20)	(0.20)	(0.23)
Demographic Controls:		Х	Х		X	x
Unemployment Controls:			X			X

TABLE 3: DID EFFECTS OF FEDERAL EITC REFORMS ON SINGLE MOTHERS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS

=

Notes: This table is constructed exactly as Table 2, but focusing on a sample of single women with low predicted earnings. Specifically, the sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE 1: FEDERAL EITC PARAMETERS



A: EITC Schedule in 2018

Notes: This figure shows federal EITC parameters for families with different numbers of EITC-eligible children. Panel A shows the EITC schedule in 2018 as a function of earnings for families with 0, 1, 2, and 3+ children. Panel B shows the maximum annual credit (in 2018 USD) for families with 0, 1, 2, and 3+ children over the period 1968-2018. The five federal EITC reforms are marked by vertical lines.

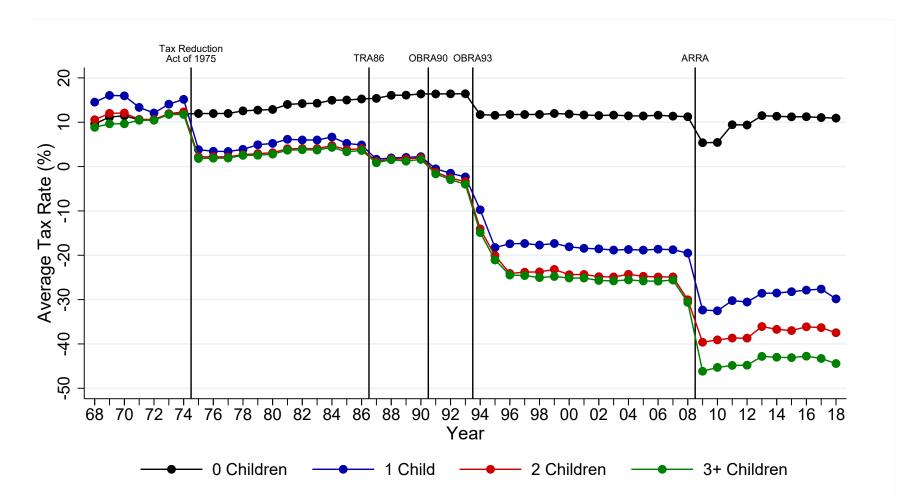
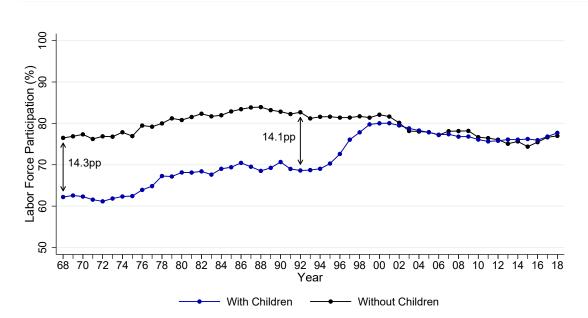


FIGURE 2: AVERAGE TAX RATES ON SINGLE WOMEN OVER TIME

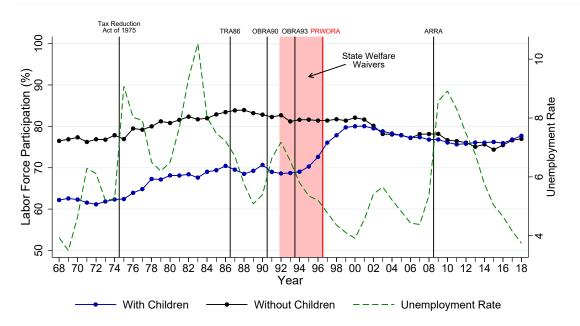
Notes: This figure shows average tax rates for single women with different numbers of EITC-eligible children over the period 1968-2018. The tax rates include all state income taxes, federal income taxes, and federal insurance contributions (FICA). The calculations assume that single women enter the labor market at the first kink of the federal EITC (for each year and number of children separately), i.e. where the EITC refund is maximized. The five federal EITC reforms are marked by vertical lines. Full details of the tax simulations are provided in section C of the online appendix.

FIGURE 3: FIFTY YEARS OF LABOR FORCE PARTICIPATION FOR SINGLE WOMEN



A: Raw Patterns

B: EITC Reforms and Confounders



Notes: This figure shows labor force participation rates of single women with and without children between 1968 and 2018. Panel B marks the timing of the five federal EITC reforms (black vertical lines), federal welfare reform (red vertical line), statewide welfare waivers (red shaded area), and the national unemployment rate (green dashed line). The outcome is weekly participation and the sample includes all single women aged 20-50 using the March CPS files.

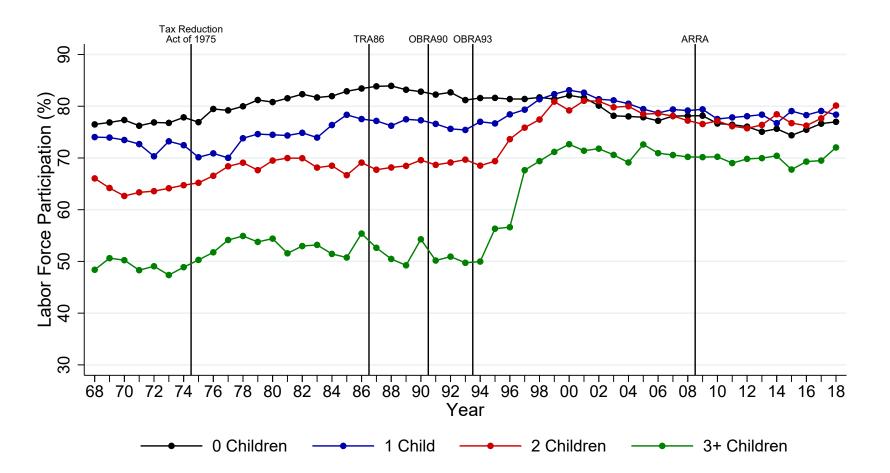
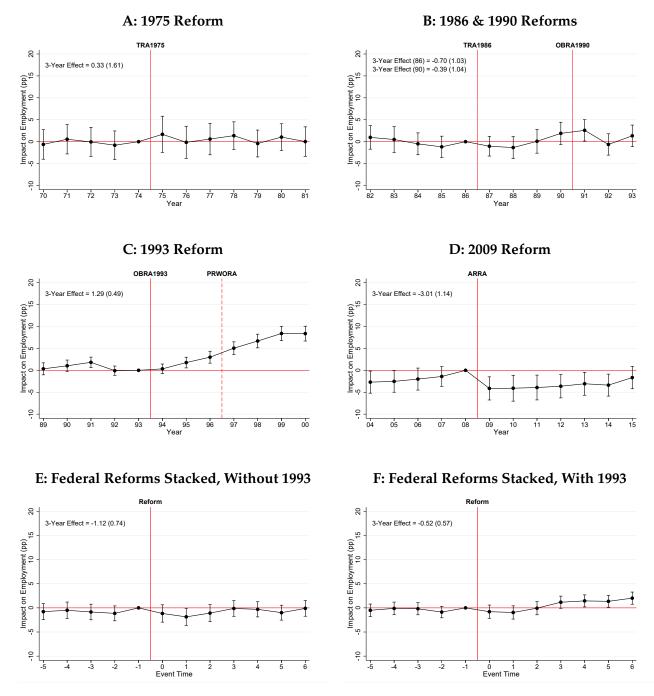


FIGURE 4: FIFTY YEARS OF LABOR FORCE PARTICIPATION FOR SINGLE WOMEN, BY NUMBER OF CHILDREN

Notes: This figure shows labor force participation rates of single women with 0, 1, 2, and 3+ children between 1968 and 2018. The timing of the five federal EITC reforms are marked by vertical lines. The outcome is weekly participation and the sample includes all single women aged 20-50 using the March CPS files.

FIGURE 5: DID EVENT STUDIES OF FEDERAL EITC REFORMS All Single Women, Weekly Employment



Notes: This figure presents DiD event studies of the five federal EITC reforms separately (Panels A-D) and stacked together (Panels E-F). Panels A-D plot estimates of the yearly impact $\hat{\gamma}_t$ based on specification (2). Panels E-F plot estimates of the average yearly impact $\bar{\gamma}_t = \sum_{k=1}^{K} \frac{\hat{\gamma}_k^k}{K}$, where $\hat{\gamma}_t^k$ is the estimated impact in event year t for reform episode k = 1, ..., K. Two stacked event studies are shown, one that excludes the confounded 1993 reform and one that includes it. The impact of each reform is based on comparing single women with and without children, except for the 2009 reform which is based on comparing single women with 3+ children to those without children. The dependent variable is weekly employment. The sample includes all single women aged 20-50. Early reforms (1975, 1986, 1990) are analyzed using the March CPS files alone, while later reforms (1993, 2009) are analyzed using the March and monthly CPS files combined. The 95% confidence intervals are based on robust standard errors clustered at the individual level. Each panel reports the average three-year DiD effect, with standard errors in parentheses.

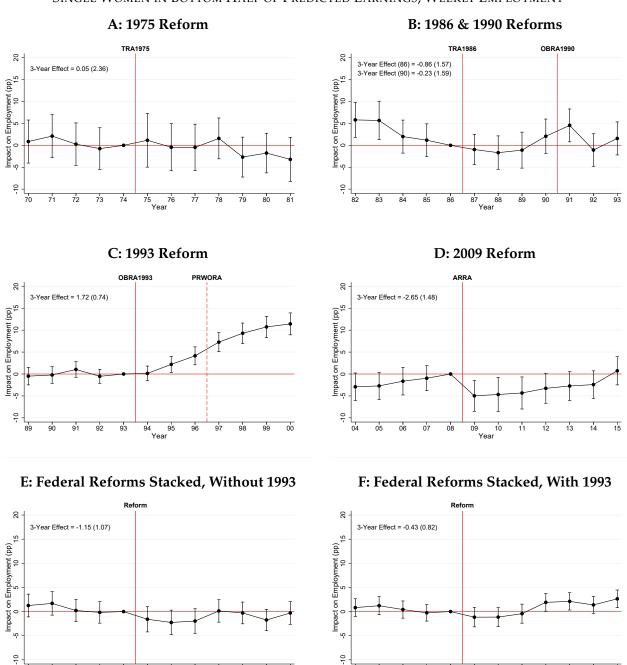


FIGURE 6: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT

Notes: This figure is constructed exactly as Figure 5, but focusing on a sample of single women with low predicted earnings. Specifically, the sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

-5 -4 -3 -2

6

5

0 1 Event Time

-1

2

3

6

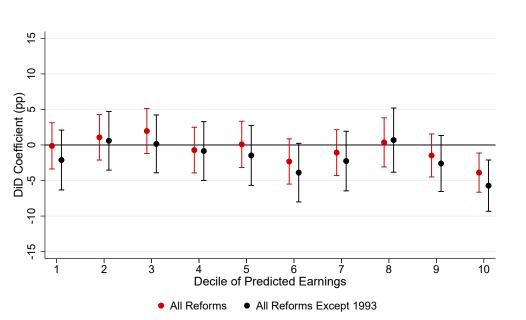
5

4

-5 -4 -3 -2 -1

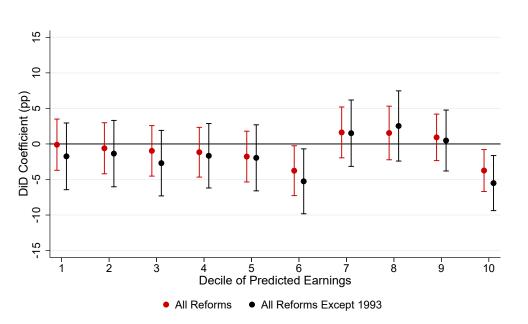
0 1 Event Time 2 3

FIGURE 7: EFFECTS OF FEDERAL REFORMS STACKED BY DECILE OF PREDICTED EARNINGS WEEKLY EMPLOYMENT

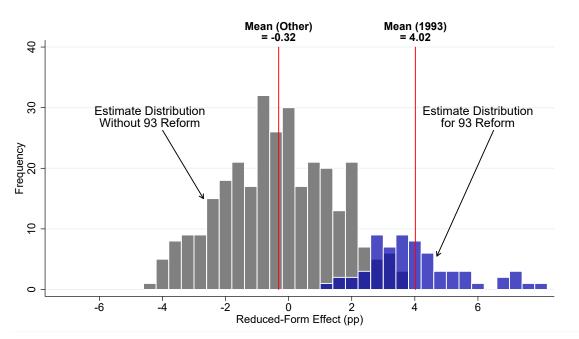


A: Raw Data

B: With Controls

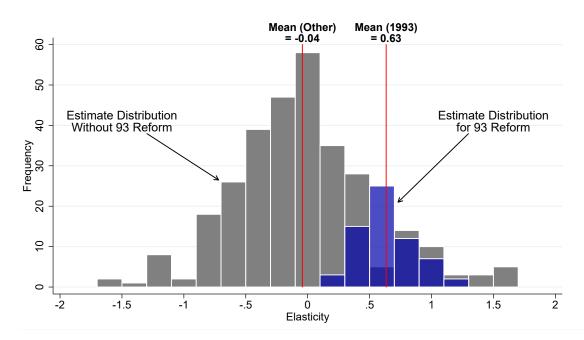


Notes: This figure shows estimates of three-year effects of federal EITC reforms by decile of predicted earnings. The estimates are based on stacked DiD specifications including either all reforms (red dots) or all reforms apart from 1993 (black dots). For each reform, predicted earnings are estimated by running specification (1) on three years of pre-reform data for working single women, using the coefficients to predict earnings for all single women in all years. The sample is then split by deciles of predicted earnings within each year. Panel A shows estimates from a specification without any controls, while Panel B adds controls for demographics and state-level unemployment rates (as in equation 2). The dependent variable is weekly employment. The sample includes all single women aged 20-50. The 95% confidence intervals are based on robust standard errors clustered at the individual level.

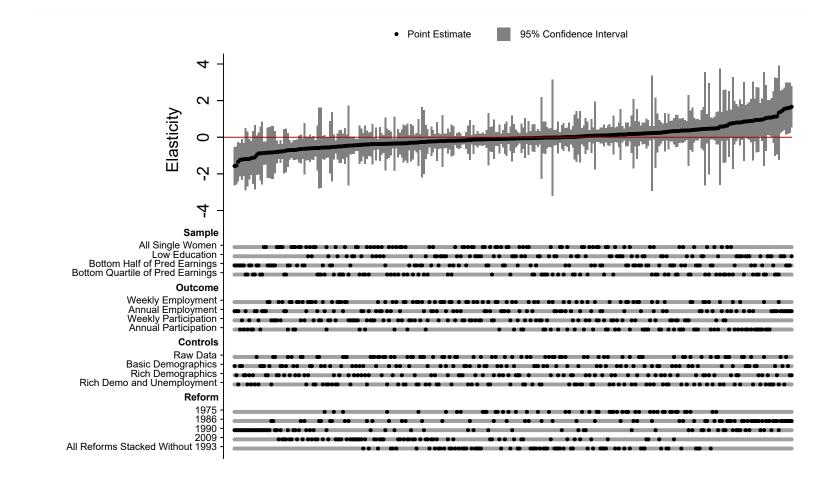


A: Distribution of Reduced-Form Estimates

B: Distribution of Elasticity Estimates



Notes: This figure shows the distribution of estimated effects of federal EITC reforms across 432 different specifications (permutations of different reform experiments, extensive margin measures, samples, and control variables). Panel A shows reduced-form effects on employment or participation rates (average three-year effects), while Panel B shows the implied extensive margin elasticities with respect to $1 - \tau$. Each panel shows two estimate distributions: all estimates excluding those based on the confounded 1993 reform (gray bars) and estimates based on the 1993 reform (blue bars). The mean of each distribution is marked by a vertical line. The distribution excluding 1993 estimates is centered around zero (with a mean elasticity of -0.04), whereas the distribution of 1993 estimates is shifted to the right (with a mean elasticity of 0.63).



Notes: This figure shows elasticity estimates for federal EITC reforms (leaving out the confounded 1993 reform) across all specifications (304 permutations of different reform experiments, extensive margin measures, samples, and control variables). The elasticity estimates are shown by black dots (ranked from low to high), with the 95% confidence intervals in gray. The underlying specification for each estimate is indicated below the graph. Only a few specifications at the top (3.6% of all specifications) yield estimates that are positive *and* statistically significant.

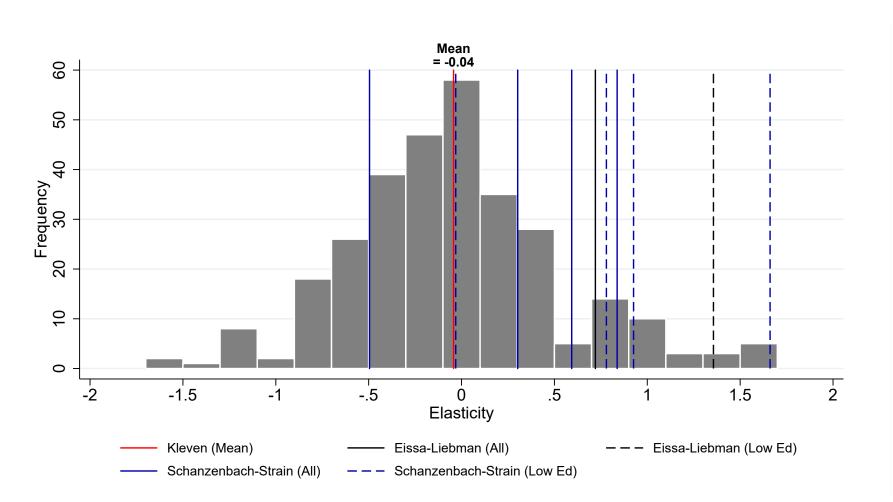
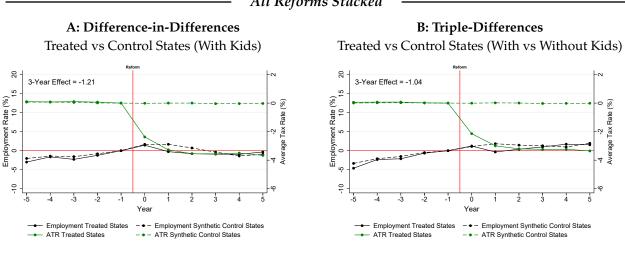


FIGURE 10: DISTRIBUTION OF ELASTICITY ESTIMATES ACROSS SPECIFICATIONS VS PRIOR ESTIMATES

Notes: This figure compares the distribution of elasticities across all specifications and reforms (leaving out the 1993 reform) to the elasticities implied by reducedform estimates for those same reforms in Eissa and Liebman (1996) (black vertical lines) and Schanzenbach and Strain (2021) (blue dashed lines). The reduced-form estimates are taken from Table III and Table IV in Eissa and Liebman (1996), and from Table 3 in Schanzenbach and Strain (2021). Both papers use annual employment as the extensive margin outcome, and they consider treatment effects on all single mothers (labeled "All" in the figure) and on low-educated single mothers (labeled "Low" in the figure). Full details on the conversion of the reduced-form estimates in these papers into elasticities can be found in Table A.4 of the online appendix.

57

FIGURE 11: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS ALL SINGLE WOMEN, WEEKLY EMPLOYMENT



All Reforms Stacked

Ten Largest Reforms Stacked

C: Difference-in-Differences Treated vs Control States (With Kids)

ò Yea

- - ATR Synthetic Control States

20

15

Employment Rate (%)

9

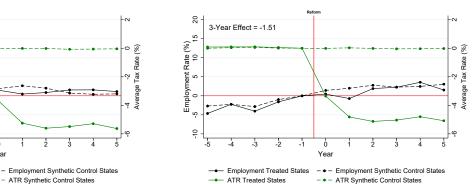
3-Year Effect = -1 25

Employment Treated States

ATR Treated States

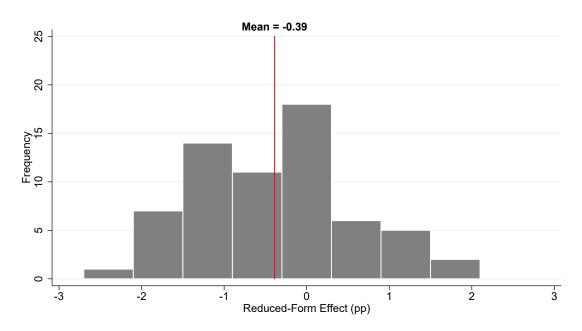
D: Triple-Differences

Treated vs Control States (With vs Without Kids)



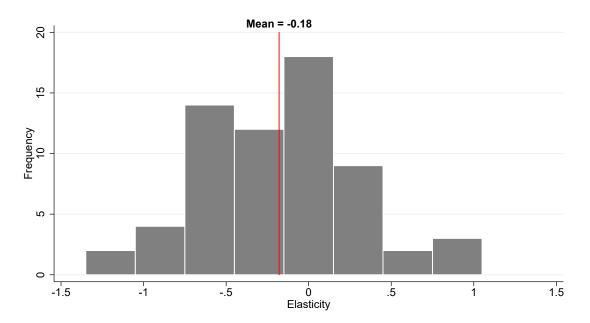
Notes: This figure shows stacked event studies of state EITC reforms using a synthetic control approach. Panels A and C are based on a difference-in-differences approach comparing states with and without EITC reforms for single women with children. Panels B and D are based on a triple-differences approach comparing states with and without EITC reforms for single women with children relative to those without children (in these panels, each series show differences between single women with and without children). The average tax rate (ATR) series include only state income taxes, and they are calculated by setting earnings equal to the first kink of the federal EITC (for each year and number of children), The top row includes all state reforms except for Iowa, Maryland, Minnesota, Rhode Island, Vermont, and Wisconsin (because of small sample sizes) and Indiana, Ohio, and Oregon (because other state tax changes offset the EITC incentive). The bottom row focuses on the ten largest state EITC reforms, i.e. those where the reform-induced ATR reduction for single women with children (relative to those without) was the largest. These are the reforms in Colorado, Connecticut, District of Columbia, Kansas, Massachusetts, Michigan, Nebraska, New Jersey, New Mexico, and New York. Reforms enacted before 1993 are analyzed based on the March CPS files alone, while reforms enacted after 1993 are analyzed based on the March and monthly CPS files combined. The sample includes all single women aged 20-50. The details of each state EITC reform are provided in appendix Table A.2, and the full details of the synthetic control approach are provided in appendix section F.

FIGURE 12: EFFECTS OF STATE EITC REFORMS ACROSS ALL SPECIFICATIONS



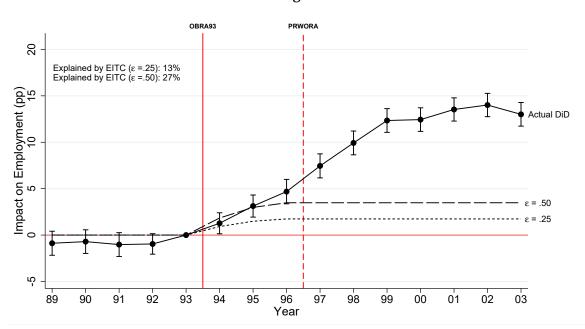
A: Distribution of Reduced-Form Estimates

B: Distribution of Elasticity Estimates



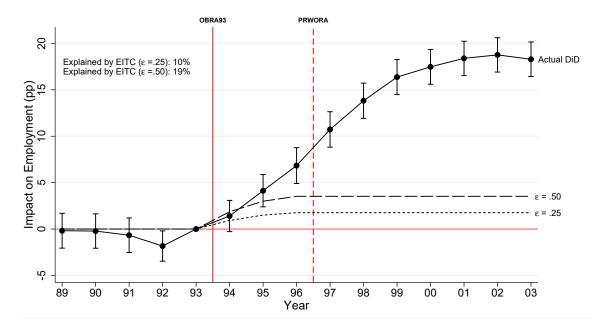
Notes: This figure shows the distribution of estimated effects of state EITC reforms across 64 different specifications. These are permutations of different reform experiments (all reforms stacked and ten largest reforms stacked), empirical strategies (difference-in-differences and triple-differences), extensive margin measures, and samples. Panel A shows reduced-form effects on employment or participation rates (average three-year effects), while Panel B shows the implied extensive margin elasticities with respect to $1 - \tau$. Reforms enacted before 1993 are analyzed based on the March CPS files alone, while reforms enacted after 1993 are analyzed based on the March and monthly CPS files combined. The sample includes single women aged 20-50.

FIGURE 13: ESTIMATED VS SIMULATED EFFECTS OF THE 1993 EITC REFORM WEEKLY EMPLOYMENT



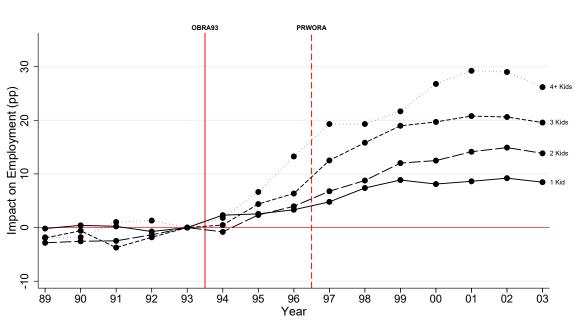
A: All Single Women

B: Single Women in Bottom Half of Predicted Earnings



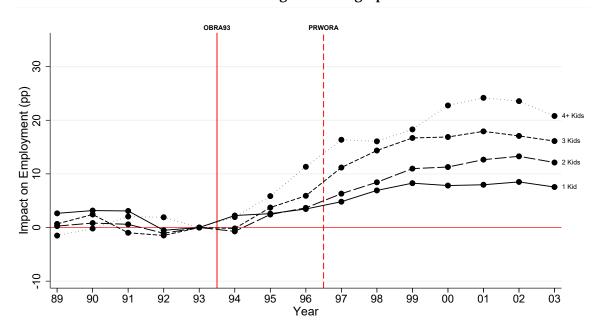
Notes: This figure shows actual and simulated DiD event studies of the 1993 reform. The actual DiD series plot estimates $\hat{\gamma}_t$ from specification (2) without any controls. The simulated DiD series plot ΔP_t calculated from equation (4), assuming an elasticity of 0.25 (short dashes) or 0.5 (long dashes). Panel A shows results for all single women, while Panel B shows results for single women in the bottom half of predicted earnings, estimated based on specification (1). Each panel also shows the fraction of the employment increase over the post-reform period that can be explained by the 1993 tax reform (including but not limited to the EITC expansion) under the two elasticity scenarios. Among the most treated women (Panel B), only 10-19% of the employment increase in the 1990s can be attributed to the EITC even under the sizable tax elasticities assumed. The extensive margin outcome is weekly employment, and the sample includes single women aged 20-50 using the March and monthly CPS files combined. The 95% confidence intervals are based on robust standard errors clustered at the individual level.

FIGURE 14: FANNING-OUT BY NUMBER OF CHILDREN All Single Women, Weekly Employment





B: Controlling for Demographics

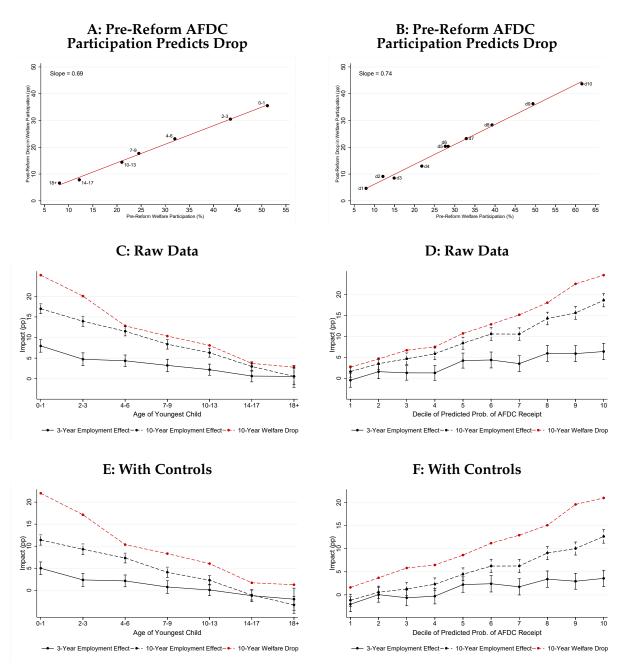


Notes: This figure shows DiD event studies of the 1993 reform by number of EITC-eligible children (1, 2, 3, 4+). The graphs plot estimates $\hat{\gamma}_t^n$ from an extension of specification (2) with separate dummies for each number of children *n*. Hence, each series shows the difference between single women with *n* children and single women without children, normalized to zero in 1993. Panel A shows raw estimates, while panel B controls for demographics. The extensive margin outcome is weekly employment. The sample includes single women aged 20-50 using the March and monthly CPS files combined.

FIGURE 15: EFFECTS OF THE 1993 EITC REFORM BY WELFARE TREATMENT INTENSITY All Single Women, Weekly Employment

By Age of Youngest Child

BY PROBABILITY OF AFDC PARTICIPATION



Notes: This figure shows DiD effects of the 1993 reform by welfare treatment intensity, estimated from equation (5). The left panels use the age of youngest child as a proxy for welfare treatment, while the right panels use deciles of a predicted probability of AFDC receipt (estimated from eq. 6) as the proxy for welfare treatment. The top row plots the level of welfare participation rates before the reform (1993) against the drop in welfare participation rates after the reform (from 1993 to 2003) across bins of the welfare treatment proxy. The next rows plot 3-year employment effects (solid black), 10-year employment effects (dashed black), and 10-year welfare drops (dashed red) by welfare treatment intensity in the raw data (middle row) and after controlling for demographics and business cycle effects (bottom row). These are average effects based on post-reform dummies for either 1994-1996 or 1994-2003 dummies. The extensive margin outcome is weekly employment. The sample includes single women aged 20-50 using the March and monthly CPS files combined. The 95% confidence intervals are based on robust standard errors clustered at the individual level.

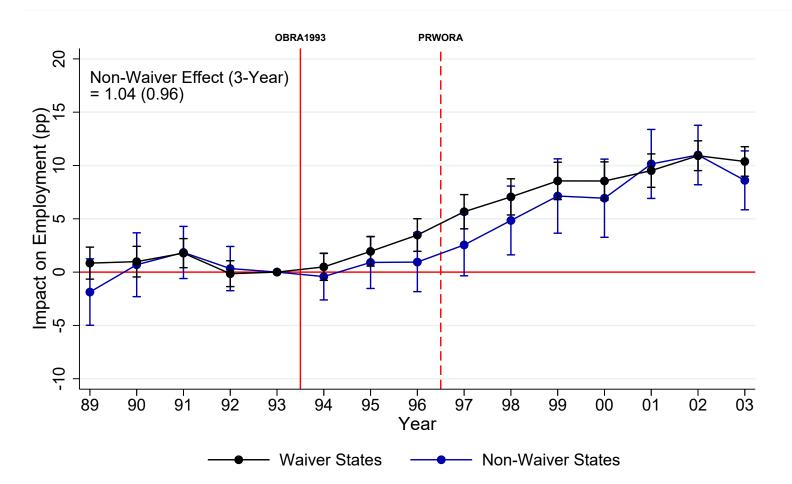


FIGURE 16: HOW MUCH CAN BE EXPLAINED BY WELFARE WAIVERS? All Single Women, Weekly Employment

Notes: This figure shows DiD event studies of the 1993 reform for waiver states (black series) and non-waiver states (blue series). Specifically, the series show estimates of the DiD coefficient γ_t from specification (2), implemented separately on states that ever approved statewide waiver legislation and those that did not. Both series include controls for demographics and unemployment. From Table A.3 in the appendix, there were 13 states without any statewide waiver legislation: Alabama, Alaska, District of Columbia, Kansas, Kentucky, Louisiana, Nevada, New Mexico, New York, Oklahoma, Pennsylvania, Rhode Island, and Wyoming. The extensive margin outcome is weekly employment. The sample includes single women aged 20-50 using the March and monthly CPS files combined. The 95% confidence intervals are based on robust standard errors clustered at the individual level.

Online Appendix (Not for Publication)

A Supplementary Figures and Tables

Year	Family	Phase-in	First	Max	Second	Phase-out	Exhaustion
	Size	Rate (%)	Kink	Credit	Kink	Rate (%)	Point
1975-78	all	10	4,000	400	4,000	10.00	8,000
1979-84	all	10	5,000	500	6,000	12.50	10,000
1985-86	all	11	5,000	550	6,500	12.22	11,000
1987	all	14	6,080	851	6,920	10.00	15,432
1988	all	14	6,240	874	9,840	10.00	18,576
1989	all	14	6,500	910	10,240	10.00	19,340
1990	all	14	6,810	953	10,730	10.00	20,264
1991	1 child	16.7	7,140	1,192	11,250	11.93	21,250
	2+ children	17.3	7,140	1,235	11,250	12.36	21,250
1992	1 child	17.6	7,520	1,324	11,840	12.57	22,370
	2+ children	18.4	7,520	1,384	11,840	13.14	22,370
1993	1 child	18.5	7,750	1,434	12,200	13.21	23,050
	2+ children	19.5	7,750	1,511	12,200	13.93	23,050
1994	0 children	7.65	4,000	306	5,000	7.65	9,000
	1 child	26.3	7,750	2,038	11,000	15.98	23,755
	2+ children	30	8,425	2,528	11,000	17.68	25,296
1995	0 children	7.65	4,100	314	5,130	7.65	9,230
	1 child	34	6,160	2,094	11,290	15.98	24,396
	2+ children	36	8,640	3,110	11,290	20.22	26,673
1996	0 children	7.65	4,220	323	5,280	7.65	9,500
	1 child	34	6,330	2,152	11,610	15.98	25,078
	2+ children	40	8,890	3,556	11,610	21.06	28,495
1997	0 children	7.65	4,340	332	5,430	7.65	9,770
	1 child	34	6,500	2,210	11,930	15.98	25,750
	2+ children	40	9,140	3,656	11,930	21.06	29,290
1998	0 children	7.65	4,460	341	5,570	7.65	10,030
	1 child	34	6,680	2,271	12,260	15.98	26,473
	2+ children	40	9,390	3,756	12,260	21.06	30,095
1999	0 children	7.65	4,530	347	5,670	7.65	10,200
	1 child	34	6,800	2,312	12,460	15.98	26,928
	2+ children	40	9,540	3,816	12,460	21.06	30,580

 TABLE A.1: FEDERAL EITC PARAMETERS OVER TIME

	Family	Phase-in	First	Max	Second	Phase-out	Exhaustion
Year	Size	Rate (%)	Kink	Credit	Kink	Rate (%)	Point
2000	0 children	7.65	4,610	353	5,770	7.65	10,380
	1 child	34	6,920	2,353	12,690	15.98	27,413
	2+ children	40	9,720	3,888	12,690	21.06	31,152
2001	0 children	7.65	4,760	364	5,950	7.65	10,710
	1 child	34	7,140	2,428	13,090	15.98	28,281
	2+ children	40	10,020	4,008	13,090	21.06	32,121
2002	0 children	7.65	4,910	376	6,150	7.65	11,060
	1 child	34	7,370	2,506	13,520	15.98	29,201
	2+ children	40	10,350	4,140	13,520	21.06	33,178
2003	0 children	7.65	4,990	382	6,240	7.65	11,230
	1 child	34	7,490	2,547	13,730	15.98	29,666
	2+ children	40	10,510	4,204	13,730	21.06	33,692
2004	0 children	7.65	5,100	390	6,390	7.65	11,490
	1 child	34	7,660	2,604	14,040	15.98	30,338
	2+ children	40	10,750	4,300	14,040	21.06	34,458
2005	0 children	7.65	5,220	399	6,530	7.65	11,750
	1 child	34	7,830	2,662	14,370	15.98	31,030
	2+ children	40	11,000	4,400	14,370	21.06	35,263
2006	0 children	7.65	5,380	412	6,740	7.65	12,120
	1 child	34	8,080	2,747	14,810	15.98	32,001
	2+ children	40	11,340	4,536	14,810	21.06	36,348
2007	0 children	7.65	5,590	428	7,000	7.65	12,590
	1 child	34	8,390	2,853	15,390	15.98	33,241
	2+ children	40	11,790	4,716	15,390	21.06	37,783
2008	0 children	7.65	5,720	438	7,160	7.65	12,880
	1 child	34	8,580	2,917	15,740	15.98	33,995
	2+ children	40	12,060	4,824	15,740	21.06	38,646
2009	0 children	7.65	5 <i>,</i> 970	457	7,470	7.65	13,440
	1 child	34	8,950	3,043	16,420	15.98	35,463
	2 children	40	12,570	5,028	16,420	21.06	40,295
	3+ children	45	12,570	5,657	16,420	21.06	43,279
2010	0 children	7.65	5,980	457	7,480	7.65	13,460
	1 child	34	8,970	3,050	16,450	15.98	35 <i>,</i> 535
	2 children	40	12,590	5,036	16,450	21.06	40,363
	3+ children	45	12 <i>,</i> 590	5,666	16,450	21.06	43,352
2011	0 children	7.65	6,070	464	7,590	7.65	13,660

Year	Family Size	Phase-in Rate (%)	First Kink	Max Credit	Second Kink	Phase-out Rate (%)	Exhaustion Point
	1 child	34	9,100	3,094	16,690	15.98	36,052
	2 children	40	12,780	5,112	16,690	21.06	40,964
	3+ children	45	12,780	5,751	16,690	21.06	43,998
2012	0 children	7.65	6,210	475	7,770	7.65	13,980
	1 child	34	9,320	3,169	17,090	15.98	36,920
	2 children	40	13,090	5,236	17,090	21.06	41,952
	3+ children	45	13,090	5,891	17,090	21.06	45,060
2013	0 children	7.65	6,370	487	7,970	7.65	14,340
	1 child	34	9,560	3,250	17,530	15.98	37,870
	2 children	40	13,430	5,372	17,530	21.06	43,038
	3+ children	45	13,430	6,044	17,530	21.06	46,227
2014	0 children	7.65	6,480	496	8,110	7.65	14,590
	1 child	34	9,720	3,305	17,830	15.98	38,511
	2 children	40	13,650	5,460	17,830	21.06	43,756
	3+ children	45	13,650	6,143	17,830	21.06	46,997
2015	0 children	7.65	6,580	503	8,240	7.65	14,820
	1 child	34	9,880	3,359	18,110	15.98	39,131
	2 children	40	13,870	5,548	18,110	21.06	44,454
	3+ children	45	13,870	6,242	18,110	21.06	47,747
2016	0 children	7.65	6,610	506	8,270	7.65	14,880
	1 child	34	9,920	3,373	18,190	15.98	39,296
	2 children	40	13,931	5,572	18,190	21.06	44,648
	3+ children	45	13,930	6,269	18,190	21.06	47,955
2017	0 children	7.65	6 <i>,</i> 670	510	8,340	7.65	15,010
	1 child	34	10,000	3,400	18,340	15.98	39,617
	2 children	40	14,040	5,616	18,340	21.06	45,007
	3+ children	45	14,040	6,318	18,340	21.06	48,340
2018	0 children	7.65	6,780	519	8,490	7.65	15,270
	1 child	34	10,180	3,461	18,660	15.98	40,320
	2 children	40	14,290	5,716	18,660	21.06	45,802
	3+ children	45	14,290	6,431	18,660	21.06	49,194

Notes: This table shows federal EITC parameters by family size since the introduction of the program in 1975. The phase-in rate corresponds to the increase in the tax credit for each additional dollar of income. The first kink point is the minimum income needed to maximize the credit. The maximum credit is largest possible EITC amount a family can receive. The second kink point is the maximum income allowed before the credit begins to phase out. The phase-out rate is the reduction in the tax credit for each additional dollar of income above the second kink point. The exhaustion point is the income level at which the EITC is completely phased out.

				Reform	Induced	l ATR Chan	ges (pp)
	Year	% of		Any	1	2	3+
State	Instituted	Fed. Credit	Туре	Children	Child	Children	Children
Alabama	•	•		•		•	•
Alaska	•		•	•	•	•	•
Arizona			•	•	•	•	•
Arkansas	•		•	•	•	•	•
California ¹	2015	n/a	R	-0.70	-1.02	-0.40	-0.56
Colorado ²	1999	10	R	-3.05	-2.74	-3.57	-3.07
Connecticut	2011	25	R	-8.32	-6.97	-8.74	-11.09
Delaware	2006	20	NR	-0.05	-0.04	-0.10	-0.04
Dist. of Columbia	2000	25	R	-7.26	-6.53	-7.17	-8.76
Florida							
Georgia							
Hawaii	2018	20	NR				
Idaho							
Illinois	2000	5	NR	-0.98	-1.03	-1.15	-0.35
Indiana ³	1999	n/a	NR	2.36	2.31	2.53	1.84
Iowa	1990	6.5	NR	-0.19	-0.44	0.16	-0.14
Kansas	1998	10	R	-3.62	-3.05	-4.34	-3.90
Kentucky							
Louisiana	2008	3.5	R	-1.23	-1.24	-1.11	-1.42
Maine	2000	5	NR	-0.04	-0.02	-0.08	-0.02
$Maryland^4$	1987	50	NR	-1.27	-1.34	-1.18	-1.30
Massachusetts	1997	10	R	-3.35	-3.07	-3.79	-3.46
Michigan	2008	20	R	-6.43	-5.81	-6.60	-7.53
Minnesota ⁵	1991	n/a	R	-2.88	-3.08	-2.69	-2.93
Mississippi							
Missouri							
Montana							
Nebraska	2005	8	R	-3.29	-2.96	-3.71	-3.66
Nevada							
New Hampshire							
New Jersey	2000	17.5	R	-6.23	-5.75	-6.84	-6.89
New Mexico	2007	10	R	-3.54	-3.00	-3.86	-4.11
New York	1994	20	R	-6.16	-5.41	-6.78	-6.90
North Carolina ⁶	2008	5	R	-2.07	-1.98	-2.03	-2.41

TABLE A.2: STATE EITC REFORMS

				Reform-	Induced	l ATR Chan	ges (pp)
	Year	% of		Any	1	2	3+
State	Instituted	Fed. Credit	Туре	Children	Child	Children	Children
North Dakota							
Ohio	2013	10	NR	0.05	0.02	0.01	0.17
Oklahoma	2002	5	R	-1.36	-1.31	-1.21	-1.72
Oregon ⁷	1997	5	NR	0.04	-0.13	0.18	0.18
Pennsylvania							
Rhode Island ⁸	1986	25	NR	-0.22	-0.30	-0.03	-0.03
South Carolina	2018	63	NR				
South Dakota							
Tennessee							
Texas							
Utah							
Vermont	1988	28	R	-4.09	-4.11	-4.14	-3.92
Virginia	2006	20	NR	-0.02	0.01	-0.04	-0.04
Washington ⁹	2008	n/a	n/a	-0.00	-0.00	-0.00	-0.00
West Virginia							
Wisconsin ¹⁰	1989	5/25/75	R	-4.30	-0.56	-4.04	-13.38
Wyoming	•						

Notes: This table shows details on state EITC reforms. The first three columns show the years in which state EITC supplements were instituted, along with their parameters 3 years after introduction. The notation R and NR refers to whether the credit is refundable or non-refundable. The last four columns show average tax rate changes induced by state EITC reforms for single women with children relative to single women with no children (allowing for a 3-year phase-in). The average tax rates in this table include only state taxes and are calculated by setting earnings equal to the first kink of the federal EITC.

1. California's EITC is not a percentage of the federal EITC, but is based on an independent schedule similar in structure to the federal schedule. In 2018, the maximum California credit was equal to 45 percent of the corresponding maximum federal credit for families with 0, 1, 2, and 3+ children.

2. Colorado's original EITC was contingent upon the state having surplus revenue. In 2015, legislation was enacted that made the credit permanent. Before 2015, it was only paid out between 1999 and 2001.

3. Until 2002, Indiana's EITC was not a percentage of the federal EITC, but was based on an independent schedule similar in structure to the federal schedule. In 2003, Indiana's credit was respecified to be 6 percent of the federal credit and became refundable.

4. Maryland also offers a 50% non-refundable credit that taxpayers can choose in place of the refundable credit.

5. Minnesota's EITC is not a percentage of the federal EITC, but is based on an independent schedule similar in structure to the federal schedule. In 2018, the maximum Minnesota credit was equal to 25, 30, 35, and 31 percent of the maximum federal credit for families with 0, 1, 2, and 3+ children, respectively.

6. North Carolina's credit was eliminated from 2014.

7. Oregon's EITC is 11% of the federal credit for families with children under three.

8. While Rhode Island explicitly enacted a state EITC in 1986, they already had an implicit EITC from the introduction of the federal credit in 1975. This is because, at that time, the Rhode Island income tax was assessed as a percentage of the federal income tax.

9. Washington enacted a state EITC in 2008, but the credit has never been funded or paid out.

10. Wisconsin introduced a non-refundable EITC already **69**1984, which was repealed in 1986 and reinstituted in 1989 as a refundable credit. Wisconsin's credit varies by family size. The numbers shown in the table correspond to the credit for 1, 2, and 3+ eligible children, respectively.

State	Termination Time Limits			equirement e Limits		BS ptions	2	BS tions		nily 1ps		nings egard
	Appr	Impl	Appr	Impl	Appr	Impl	Appr	Impl	Appr	Impl	Appr	Impl
Alabama												
Alaska												
Arizona	5-95	11-95				11-95	5-95	11-95	5-95	11-95		
Arkansas									4-94	7-94		
California			9-95	9-95					8-96	9-97	10-92	12-92
Colorado												6-94
Connecticut	12-95	1-96			8-94	1-96	8-94	1-96	12-95	1-96	8-94	1-96
Delaware	5-95	10-95	5-95	10-95	5-95	10-95	5-95	10-95	5-95	10-95	5-95	10-95
Dist. of Columbia												
Florida		2-94			6-96				6-96			2-94
Georgia							11-93	1-94	11-93	1-94	6-94	
Hawaii	8-96	2-97			6-94	2-97					8-96	2-97
Idaho					8-96		8-96					
Illinois					9-95		9-95	10-95	9-95	12-95	11-93	11-93
Indiana	12-94	5-95			12-94	5-95	12-94	5-95	12-94	5-95		
Iowa	8-93	10-93			8-93	10-93	8-93	10-93			8-93	10-93
Kansas												
Kentucky												
Louisiana												
Maine					6-96							
Maryland					8-96	10-96	8-96	10-96	8-95	3-96	8-96	10-96
Massachusetts			8-95	11-95	8-95	11-95	8-95	11-95	8-95	11-95	8-95	11-95
Michigan			8-92		10-94	10-94	10-94	10-94			8-92	10-92
Minnesota												4-94
Mississippi									9-95	10-95		10-95
Missouri			4-95				4-95	6-95			1-93	7-94

TABLE A.3: APPROVAL AND IMPLEMENTATION DATES OF STATEWIDE WAIVERS

State		nation Limits	Work Requirement Time Limits		JO Exem	BS ptions	JO Sanc	BS tions	Family Caps			ings egard
	Appr	Impl	Appr	Impl	Appr	Impl	Appr	Impl	Appr	Impl	Appr	Impl
Montana			4-95	2-96	4-95	2-96	4-95	2-96			4-95	2-96
Nebraska	2-95	10-95			2-95	10-95	2-95	10-95	2-95	10-95	2-95	
Nevada												
New Hampshire			6-96		6-96		6-96				6-96	
New Jersey					7-92	10-92	7-92	10-92	7-92	10-92	7-92	
New Mexico												
New York												
North Carolina	2-96	7-96			2-96	7-96	2-96	7-96	2-96	7-96		
North Dakota								7-96				10-96
Ohio	3-96						3-96	7-96			3-96	7-96
Oklahoma												
Oregon	3-96	7-96			7-92	2-93	7-92	2-93				
Pennsylvania												
Rhode Island												
South Carolina	5-96				5-96		5-96		5-96			
South Dakota			3-94	6-94			3-94	6-94				
Tennessee	7-96	10-96			7-96	9-96	7-96	9-96	7-96	9-96	7-96	9-96
Texas	3-96	6-96			3-96	6-96	3-96	6-96				
Utah					10-92	1-93	10-92	1-93			10-92	1-93
Vermont			4-93	7-94	4-93	7-94	4-93	7-94			4-93	7-94
Virginia	7-95	7-95	7-95	7-95	7-95	7-95	7-95	7-95	7-95	7-95	7-95	7-95
Washington	9-95	1-96										
West Virginia							7-95	2-96				
Wisconsin			9-96	9-96	8-95	1-96	8-95	1-96	6-94	1-96		
Wyoming												

Source: Table B in Department of Health and Human Services (1999). Footnotes to Table B are used to modify certain cells of the table.

Notes: This table shows dates of approval and implementation for the six main types of statewide welfare waivers. For waivers that were rolled out at the county level, dates of implementation correspond to the date the first county implemented the reform.

Sample	Reform	Employment Effect (pp)	$\frac{\Delta P}{P}$	$\frac{\Delta(1-\tau)}{1-\tau}$	Elasticity
Panel A: Eissa and Liebman (1996)					
All Single Mothers	1986	0.022	0.030	0.042	0.71
Low-Educated Single Mothers	1986	0.038	0.057	0.043	1.34
Panel B: Schanzenbach and Strain (2021)					
All Single Mothers	1975	0.067	0.103	0.124	0.84
Low-Educated Single Mothers	1975	0.071	0.115	0.124	0.93
All Single Mothers	1986	0.014	0.021	0.035	0.59
Low-Educated Single Mothers	1986	0.037	0.060	0.036	1.66
All Single Mothers	1990	0.009	0.013	0.042	0.30
Low-Educated Single Mothers	1990	0.021	0.032	0.042	0.78
All Single Mothers	2009	-0.015	-0.022	0.044	-0.50
Low-Educated Single Mothers	2009	-0.001	-0.001	0.043	-0.03

TABLE A.4: CONVERTING PRIOR ESTIMATES INTO ELASTICITIES

Notes: This table converts reduced-form estimates from Eissa and Liebman (1996) and Schanzenbach and Strain (2021) into elasticities with respect to $1 - \tau$. These papers estimate the effects of federal EITC reforms on annual employment for all single mothers and low-educated single mothers, respectively. For Eissa and Liebman (1996), I use their estimates reported in Table III (column 5) for all single women and Table IV (columns 1 and 2) for low-educated single women. As they report separate estimates for those below high school and those at the high school level, the Eissa-Liebman estimate for low-educated single mothers shown above is a weighted average of their two estimates using the pre-reform (1984-1986) fractions of low-educated single mothers at each level. The Eissa-Liebman estimates represent four-year effects of the 1986 reform (comparing 1984-1986 to 1988-1990). For Schanzenbach and Strain (2021), I use their estimates reported in Table 3 (column 2). These are average three-year effects in the sample of either all single women or low-educated single women, defined as those with high school or less. I focus on their estimates for the 1975, 1986, 1990, and 2009 reforms, leaving out the confounded 1993 reform. Both papers give estimates of the effect on the employment rate ΔP (shown in the second column above). When calculating the percentage change in the employment rate $\Delta P/P$ and in the average net-of-tax rate $\Delta(1-\tau) / (1-\tau)$, I replicate as closely as possible the estimations samples of each paper. The change in the average net-of-tax rate in the denominator of the elasticity is calculated to correspond to the reduced-form estimate in the numerator of the elasticity. Specifically, it is a difference-in-differences calculation comparing treatment and control groups over the same time horizon as the reduced-form effect: the total change between 1984-1986 and 1988-1990 for Eissa-Liebman and the average change over the first three post-reform years for Schanzenbach-Strain. This tax term is based on the tax rate series in Figure 2. These tax rates account for all income and payroll tax provisions, and they assume that single women enter the labor market at the first kink of the federal EITC (for each year and family size).

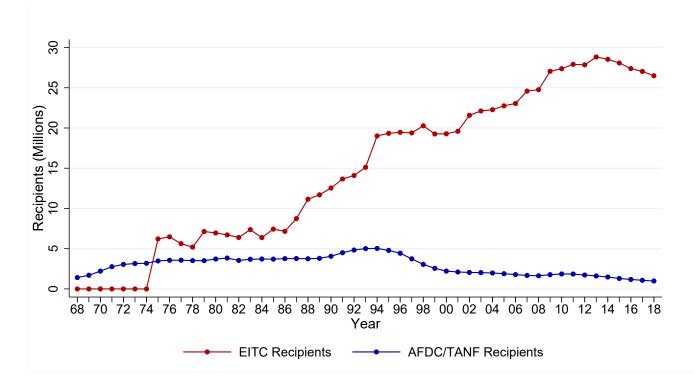
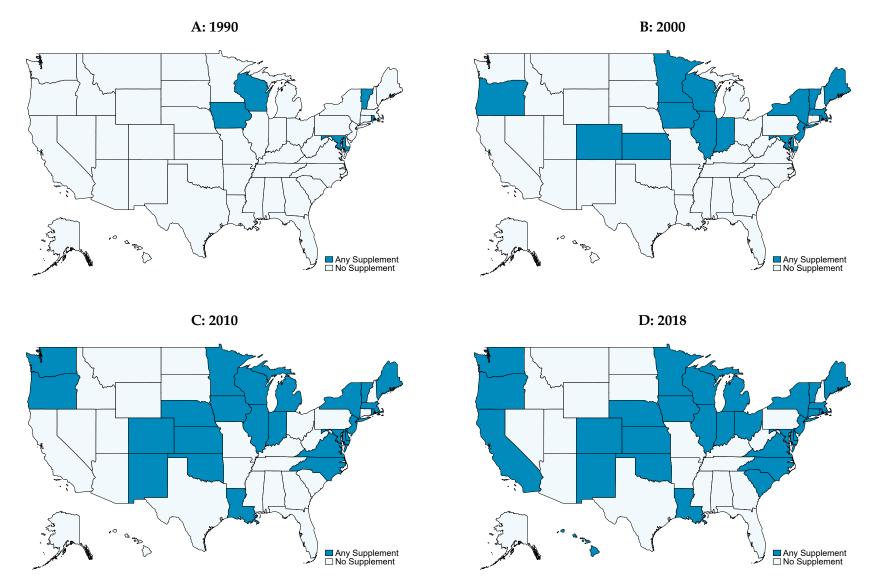


FIGURE A.1: LONG-RUN EVOLUTION OF EITC AND CASH WELFARE RECIPIENTS

Source: Internal Revenue Service (EITC) and Department of Health and Human Services (AFDC/TANF).

Notes: The red series show the number of families receiving the federal EITC between 1968-2018. The blue series show the number of families receiving Aid to Families with Dependent Children (AFDC) between 1968-1996 and Temporary Assistance for Needy Families (TANF) between 1997-2018.





Notes: The maps show which states had an EITC supplement (dark blue) and which states did not (light blue) in 1990, 2000, 2010, and 2018.

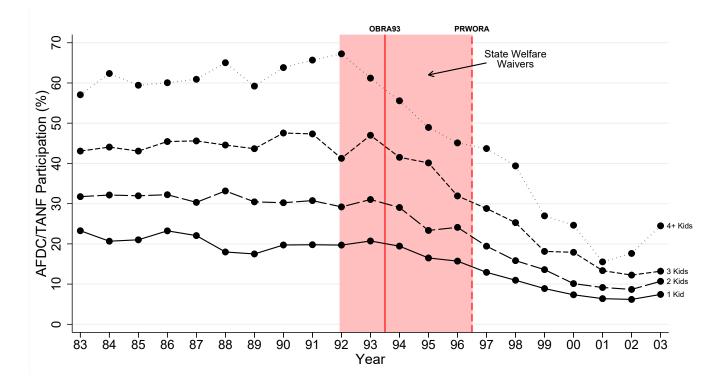
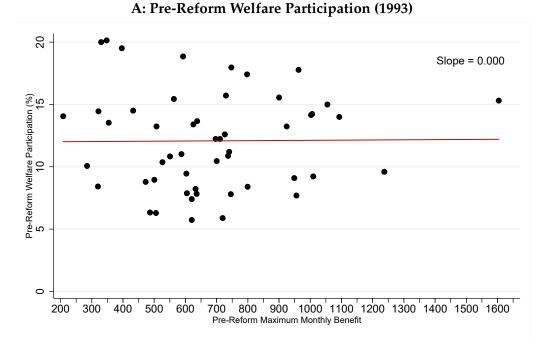
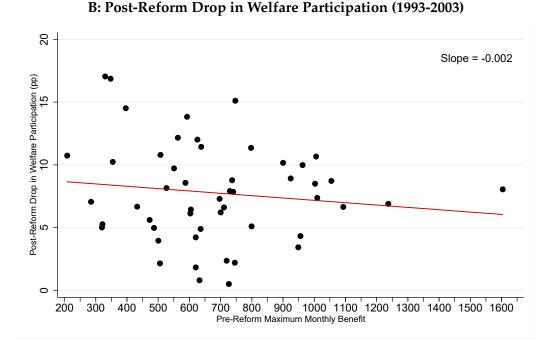


FIGURE A.3: AFDC/TANF PARTICIPATION RATES FOR SINGLE MOTHERS BY NUMBER OF CHILDREN

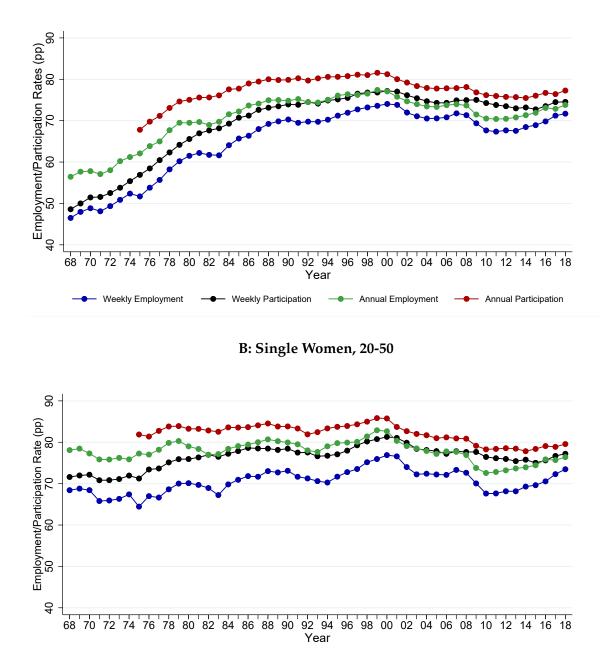
Notes: This figure shows AFDC/TANF participation rates for single mothers by number of EITC-eligible children (1, 2, 3, 4+) over the period 1983-2003. The 1993 federal EITC reform is marked by the solid red line, the 1996 federal welfare reform is marked by the dashed red line, and statewide welfare waiver reforms are marked by the red shaded area.





Notes: This figure investigates the empirical relationship between welfare participation rates and welfare benefit levels across US states in the 1990s. Panel A plots pre-reform welfare participation rates against pre-reform maximum monthly benefits across states, while Panel B plots post-reform drops in welfare participation rates against pre-reform maximum monthly benefits across states. The pre-reform year (1993) is before both EITC and AFDC/TANF reforms, while the post-reform period (1993-2003) includes both EITC and AFDC/TANF reforms. The maximum monthly benefit levels (measured in 2018 USD) are those applying to single women with two children. The sample includes single women aged 20-50 using the March files. The figure shows that state variation in pre-reform drops in welfare participation rates (Panel A), and therefore does not predict post-reform drops in welfare participation rates either (Panel B). This absence of a first stage implies that state variation in welfare benefit levels is not useful for estimating the impact of welfare reform in the 1990s.





A: All Women, 20-50

Notes: This figure shows the evolution of weekly employment, weekly participation, annual employment, and annual participation between 1968-2018. Panel A considers all women (panel A) aged 20-50, while Panel B considers single women aged 20-50. The samples are based on the March CPS files. See section B.1 for additional details.

Annual Employment

Annual Participation

Weekly Participation

Weekly Employment

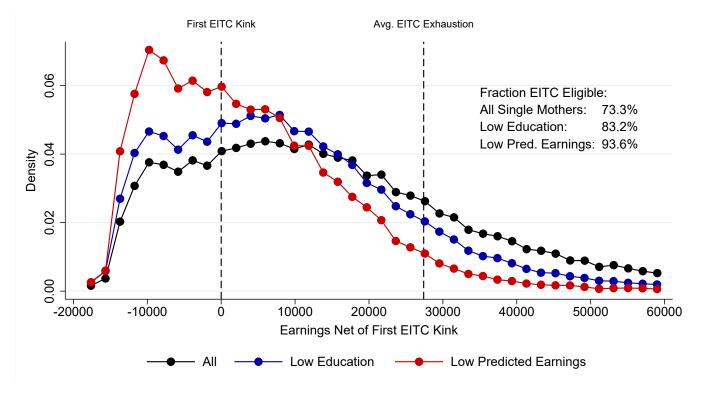
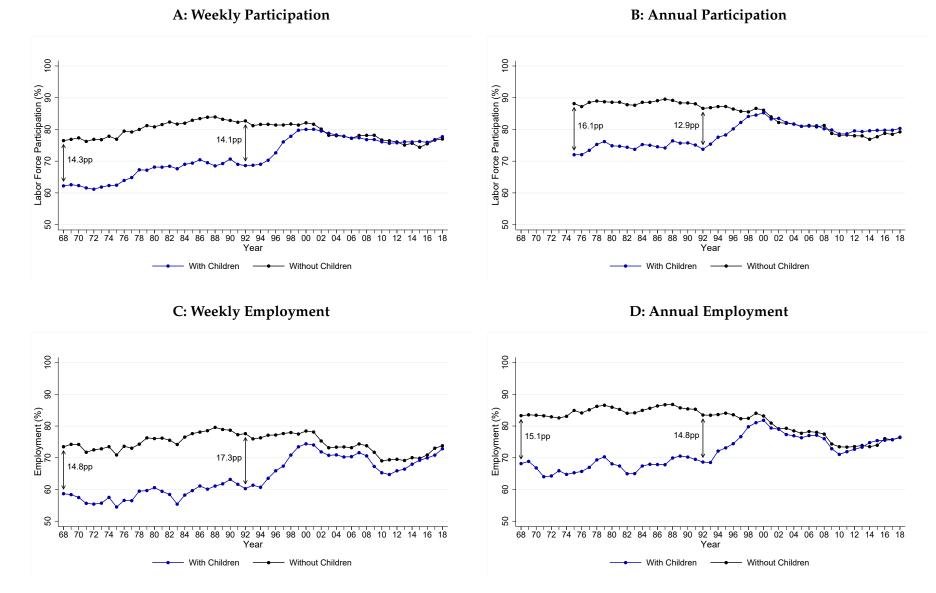


FIGURE A.6: EARNINGS DISTRIBUTIONS FOR SINGLE MOTHERS

Notes: This figure shows earnings distributions in different samples of single mothers. The distributions are based on pooled earnings data from 1975-2018, where earnings in each year are measured relative to the first EITC kink in that year. Hence, having earnings at the first EITC kink corresponds to zero. The average EITC exhaustion point (again, relative to the first kink) is calculated as an average across all years and observations. The distributions are shown for all single mothers, low-educated single mothers, and single mothers with low predicted earnings. Low-educated single mothers are those with a high school degree or less, while low-earning single mothers are those in the bottom quartile of the within-year distribution of predicted earnings, estimated from equation (1). For each of the three samples, the figure reports the fraction who are EITC eligible (i.e., below the EITC exhaustion point). The samples include single mothers aged 20-50 using the March CPS files. All dollar values have been inflation adjusted to 2018.

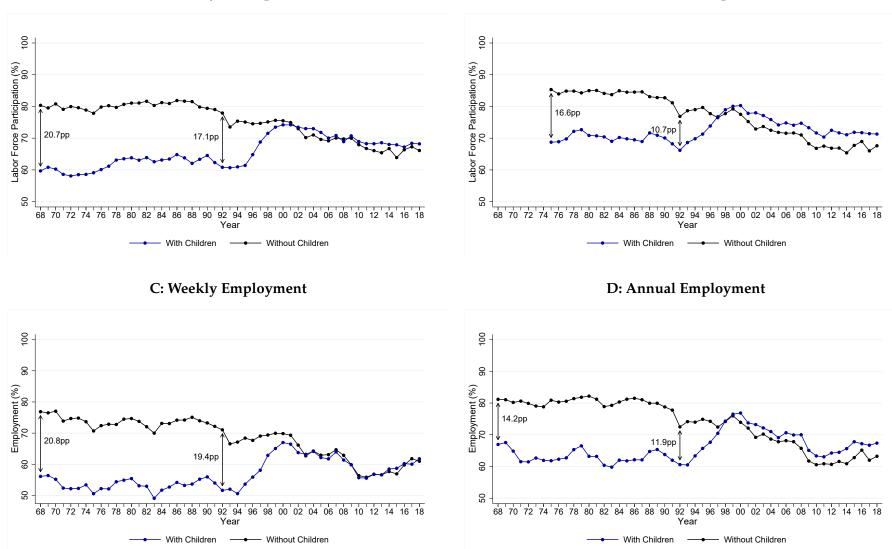
FIGURE A.7: FIFTY YEARS OF LABOR FORCE PARTICIPATION FOR SINGLE WOMEN DIFFERENT EXTENSIVE MARGIN MEASURES, ALL SINGLE WOMEN



Notes: This figure shows the weekly participation, annual participation, weekly employment, and annual employment rates of single women with and without children between 1968 and 2018. The sample includes all single women aged 20-50 using the March CPS files.

FIGURE A.8: FIFTY YEARS OF LABOR FORCE PARTICIPATION FOR SINGLE WOMEN

DIFFERENT EXTENSIVE MARGIN MEASURES, LOW-EDUCATED SINGLE WOMEN

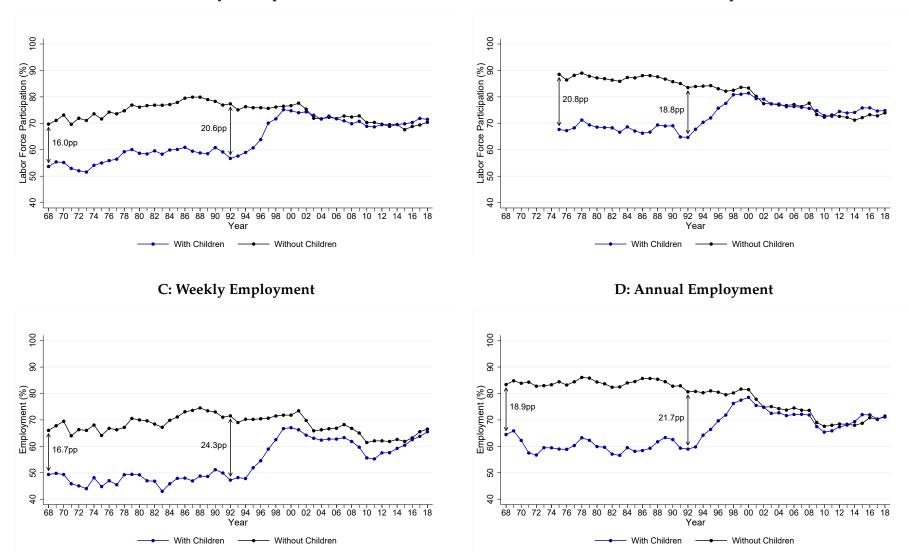


A: Weekly Participation

B: Annual Participation

Notes: This figure shows the weekly participation, annual participation, weekly employment, and annual employment rates of single women with and without children between 1968 and 2018. The sample includes low-educated single women (those with a high school degree or less) aged 20-50 using the March CPS files.

FIGURE A.9: FIFTY YEARS OF LABOR FORCE PARTICIPATION FOR SINGLE WOMEN DIFFERENT EXTENSIVE MARGIN MEASURES, SINGLE WOMEN IN THE BOTTOM HALF OF PREDICTED EARNINGS

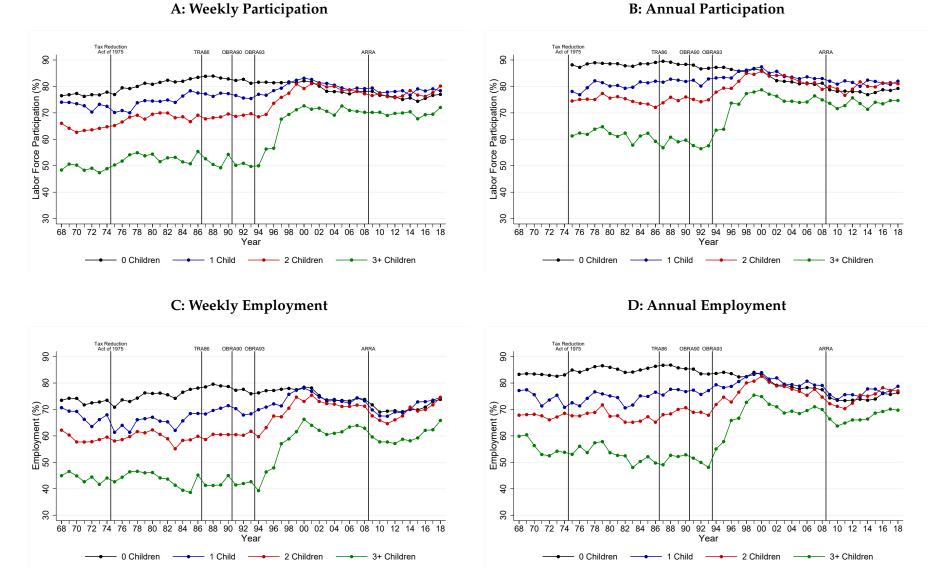


A: Weekly Participation

B: Annual Participation

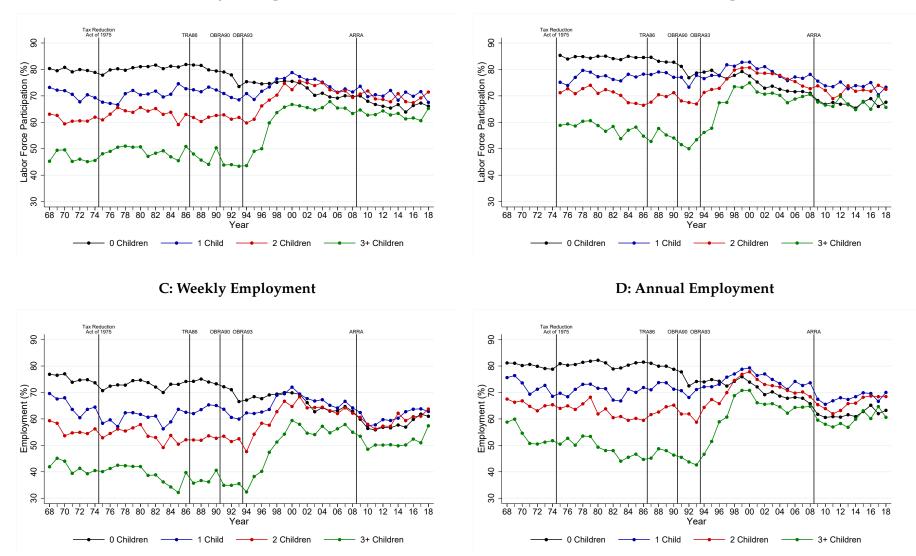
Notes: This figure shows the weekly participation, annual participation, weekly employment, and annual employment rates of single women with and without children between 1968 and 2018. The sample includes single women with low predicted earnings aged 20-50 using the March CPS files. The low earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.10: FIFTY YEARS OF PARTICIPATION FOR SINGLE WOMEN, BY NUMBER OF KIDS DIFFERENT EXTENSIVE MARGIN MEASURES, ALL SINGLE WOMEN



Notes: This figure shows the weekly participation, annual participation, weekly employment, and annual employment rates of single women with 0, 1, 2, and 3 or more children between 1968 and 2018. The sample includes all single women aged 20-50 using the March CPS files.

FIGURE A.11: FIFTY YEARS OF PARTICIPATION FOR SINGLE WOMEN, BY NUMBER OF KIDS DIFFERENT EXTENSIVE MARGIN MEASURES, LOW-EDUCATED SINGLE WOMEN



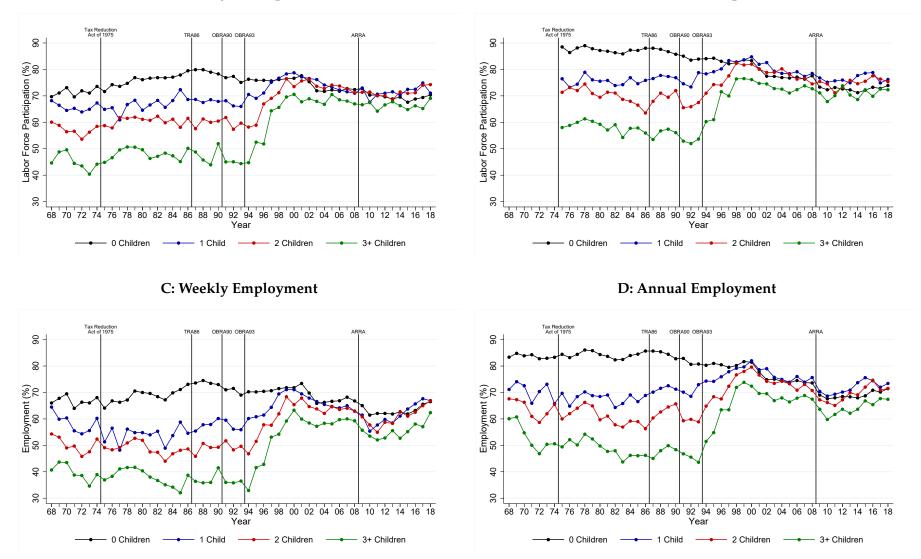
A: Weekly Participation

B: Annual Participation

Notes: This figure shows the weekly participation, annual participation, weekly employment, and annual employment rates rate of single women with 0, 1, 2, and 3 or more between 1968 and 2018. The sample includes low-educated single women (those with a high school degree or less) aged 20-50 using the March CPS files.

83

FIGURE A.12: FIFTY YEARS OF PARTICIPATION FOR SINGLE WOMEN, BY NUMBER OF KIDS DIFFERENT EXTENSIVE MARGIN MEASURES, SINGLE WOMEN IN THE BOTTOM HALF OF PREDICTED EARNINGS

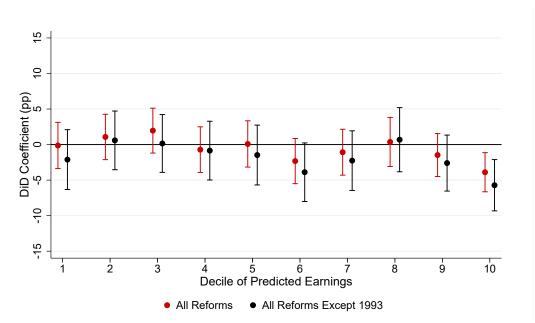


A: Weekly Participation

B: Annual Participation

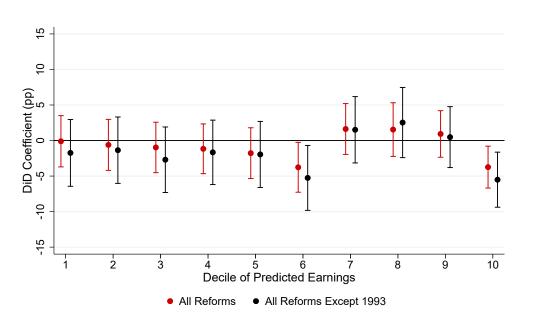
Notes: This figure shows the weekly participation, annual participation, weekly employment, and annual employment rates rate of single women with 0, 1, 2, and 3 or more between 1968 and 2018. The sample includes single women with low predicted earnings aged 20-50 using the March CPS files. The low earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.13: EFFECTS OF FEDERAL REFORMS STACKED BY DECILE OF PREDICTED EARNINGS WEEKLY EMPLOYMENT



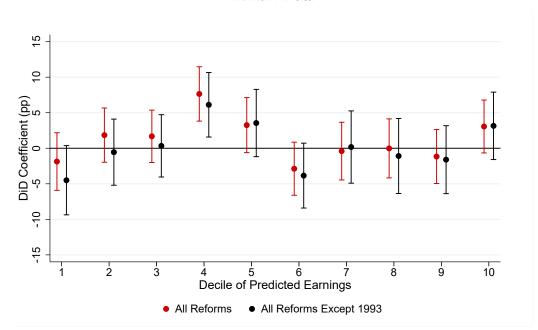
A: Raw Data

B: With Controls



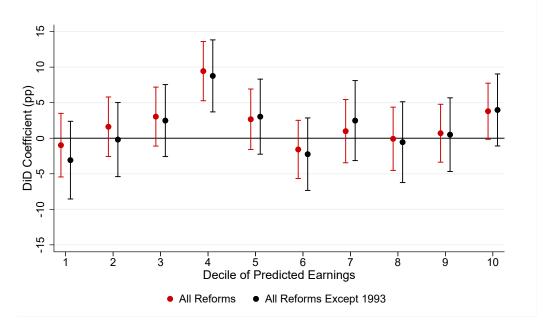
Notes: This figure is exactly the same as Figure 7 in the main text.

FIGURE A.14: EFFECTS OF FEDERAL REFORMS STACKED BY DECILE OF PREDICTED EARNINGS ANNUAL EMPLOYMENT



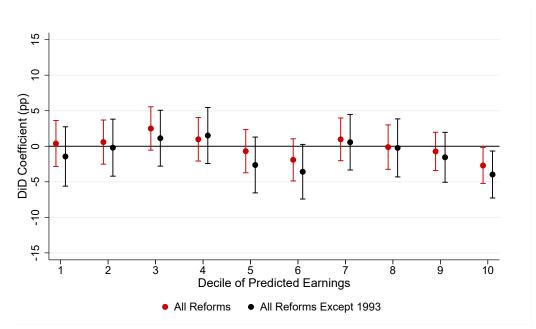
A: Raw Data

B: With Controls



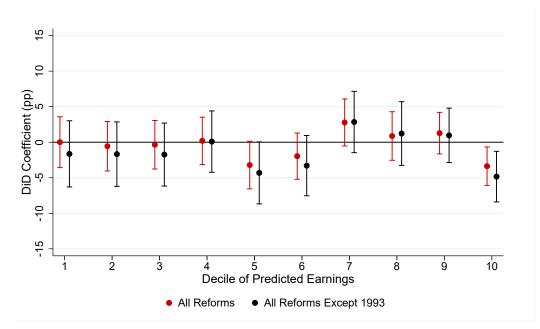
Notes: This figure is constructed exactly as Figure 7 in the main text, but using annual employment as the extensive margin measure.

FIGURE A.15: EFFECTS OF FEDERAL REFORMS STACKED BY DECILE OF PREDICTED EARNINGS WEEKLY PARTICIPATION



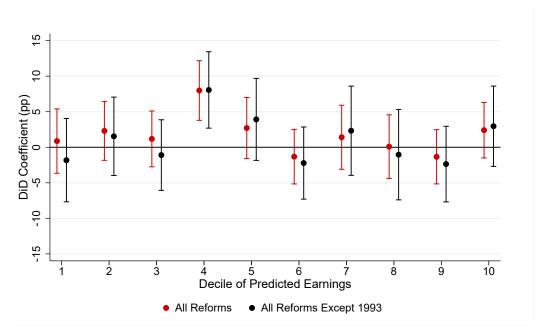
A: Raw Data

B: With Controls



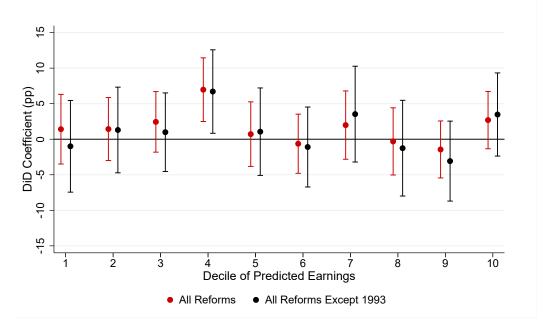
Notes: This figure is constructed exactly as Figure 7 in the main text, but using weekly participation as the extensive margin measure.

FIGURE A.16: EFFECTS OF FEDERAL REFORMS STACKED BY DECILE OF PREDICTED EARNINGS ANNUAL PARTICIPATION



A: Raw Data

B: With Controls



Notes: This figure is constructed exactly as Figure 7 in the main text, but using annual participation as the extensive margin measure.

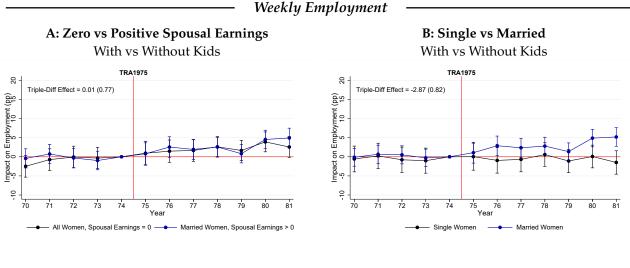
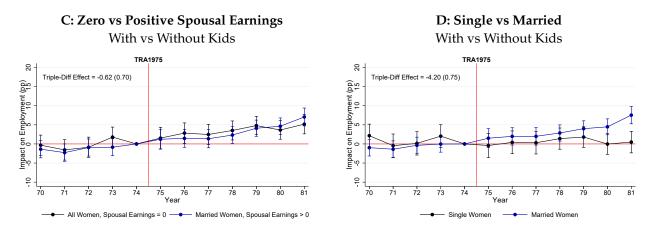


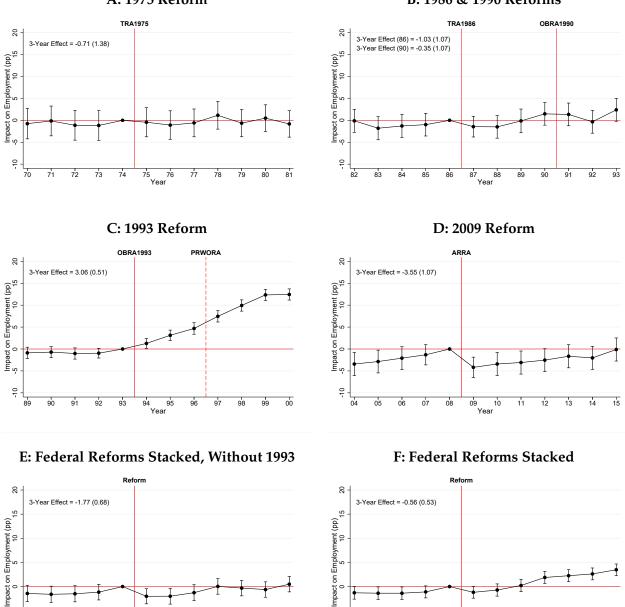
FIGURE A.17: EVENT STUDIES OF THE 1975 EITC INTRODUCTION SINGLE AND MARRIED WOMEN

Annual Employment

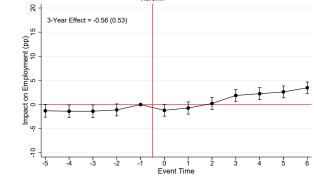


Notes: This figure presents event studies of the 1975 EITC introduction using the full sample of single and married women. The analysis is based on the triple-differences specification described in Appendix E. This specification compares treated and untreated women (in terms of their marital status and spousal earnings) with and without children over time. As described in the main text, "untreated" does not necessarily mean unaffected, but rather that the EITC incentive is either neutral or negative when accounting for the program's reliance on joint household income. Each panel depicts two difference-in-differences series (for treated and untreated women, respectively) and reports a triple-differences estimate of the average post-reform effect based on comparing the two difference-in-differences. The left panels include single women and married women with spousal earnings of zero in the treatment group (and married women in the control group), while the right panels include only single women in the treatment group (and married women in the control group). The top panels consider effects on weekly employment, while the bottom panels consider effects on annual employment. The 95% confidence intervals are based on robust standard errors. The analysis is based on March CPS files.

FIGURE A.18: DID EVENT STUDIES OF FEDERAL EITC REFORMS ALL SINGLE WOMEN, WEEKLY EMPLOYMENT RAW DATA



A: 1975 Reform B: 1986 & 1990 Reforms



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables. The figure is based on specification (2) without any controls, the sample of all single women, and using weekly employment as the extensive margin measure.

6

0 1 Event Time

2 3 4 5

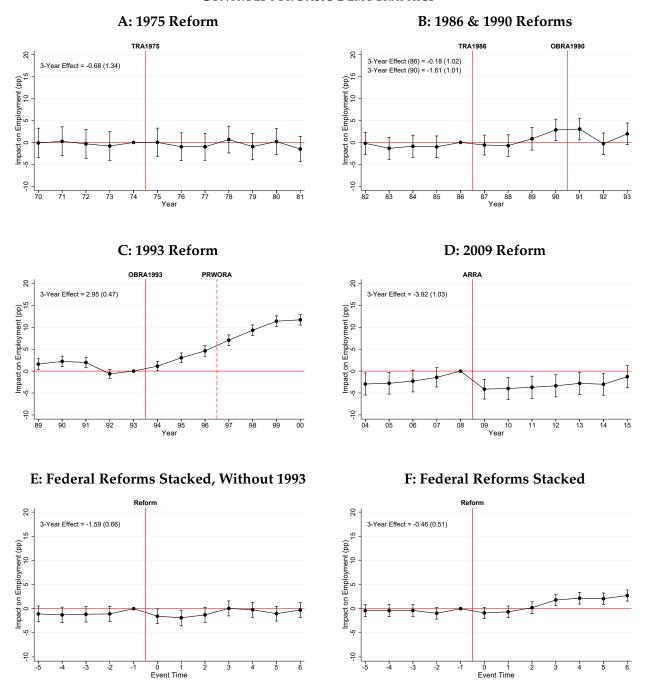
s

-10

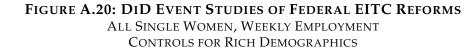
-5 4 -3 -2 -1

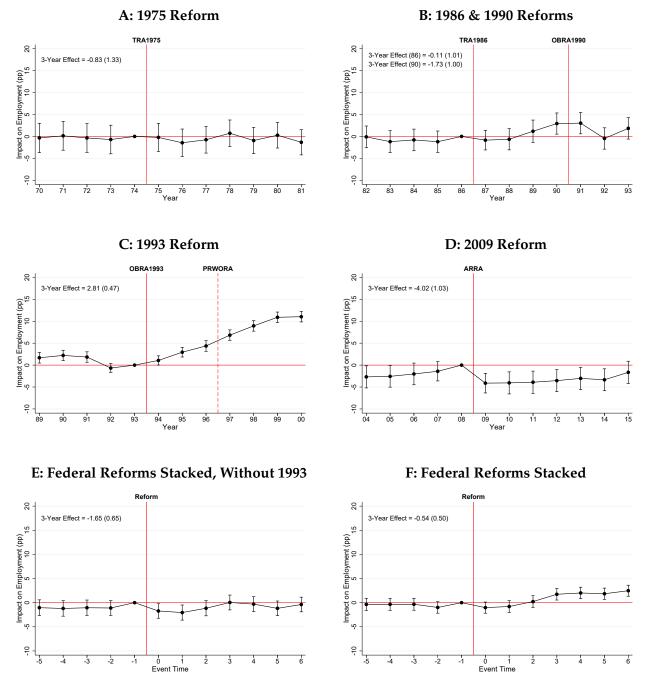
FIGURE A.19: DID EVENT STUDIES OF FEDERAL EITC REFORMS All Single Women, Weekly Employment

CONTROLS FOR BASIC DEMOGRAPHICS



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables. The figure is based on specification (2) with controls for basic demographics only, the sample of all single women, and using weekly employment as the extensive margin measure.



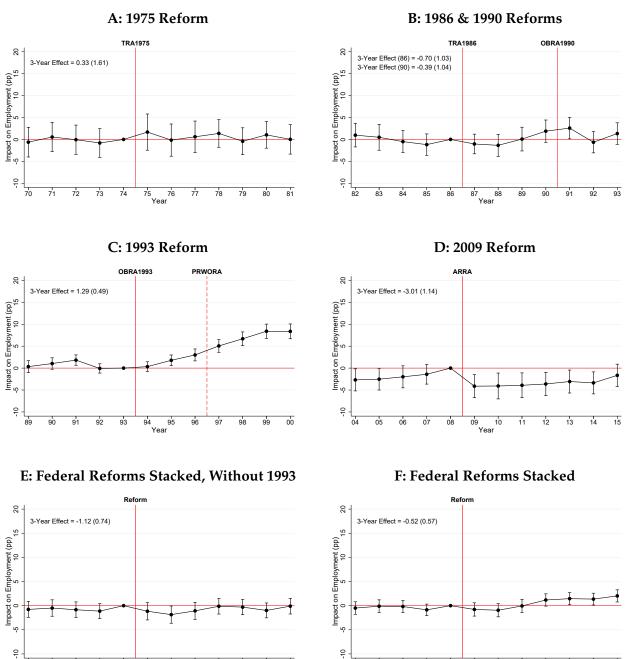


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables. The figure is based on specification (2) with controls for rich demographics only, the sample of all single women, and using weekly employment as the extensive margin measure.

FIGURE A.21: DID EVENT STUDIES OF FEDERAL EITC REFORMS

All Single Women, Weekly Employment

CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



Notes: This figure is exactly the same as Figure 5 in the main text. The figure is based on specification (2) with controls for rich demographics and unemployment, the sample of all single women, and using weekly employment as the extensive margin measure.

-5 -4 -3 -2 -1

0 1 Event Time 2 3

6

5

6

5

-5 -4 -3 -2 -1

0 1 Event Time 2 3

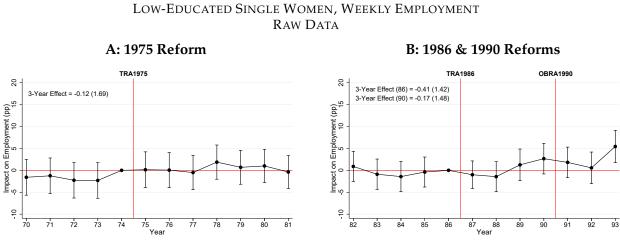
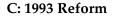
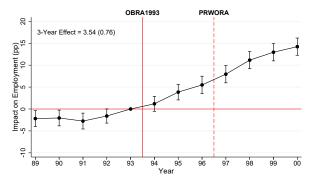


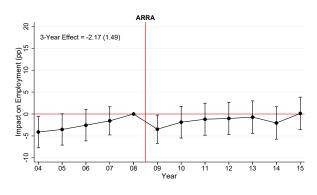
FIGURE A.22: DID EVENT STUDIES OF FEDERAL EITC REFORMS

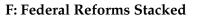


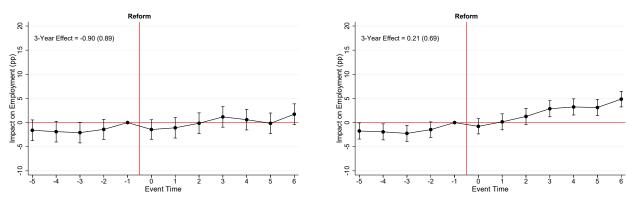


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform







Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and estimation sample. The figure is based on specification (2) without any controls, a sample of low-educated single women (those with a high school degree or less), and using weekly employment as the extensive margin measure.

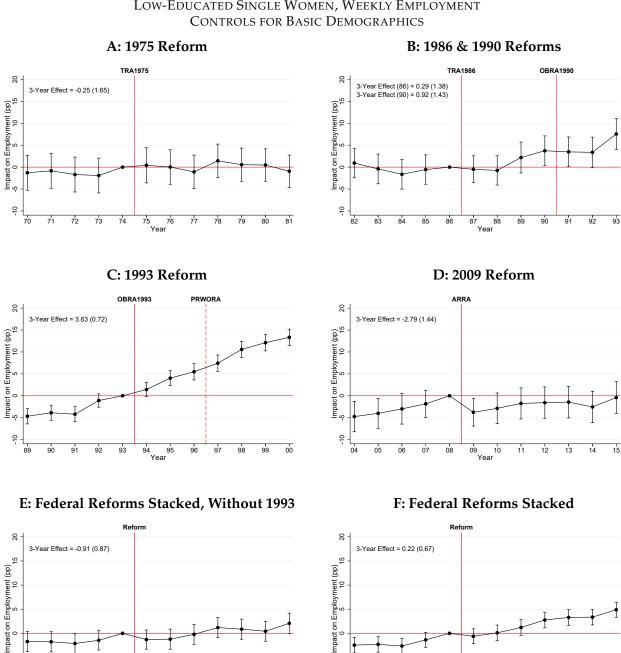


FIGURE A.23: DID EVENT STUDIES OF FEDERAL EITC REFORMS

LOW-EDUCATED SINGLE WOMEN, WEEKLY EMPLOYMENT

Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and estimation sample. The figure is based on specification (2) with controls for basic demographics only, a sample of low-educated single women (those with a high school degree or less), and using weekly employment as the extensive margin measure.

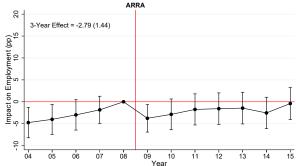
6

-10

-5

-4

-3 -2 -1



0 1 Event Time

2 3 4 5 6



0 1 Event Time

2 3 4 5

-10

-5 -4 -3 -2 -1

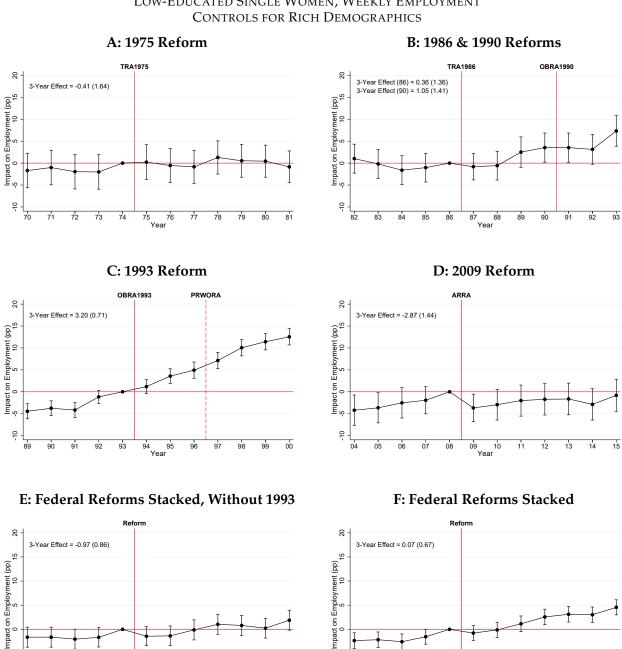


FIGURE A.24: DID EVENT STUDIES OF FEDERAL EITC REFORMS

LOW-EDUCATED SINGLE WOMEN, WEEKLY EMPLOYMENT

Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and estimation sample. The figure is based on specification (2) with controls for rich demographics only, a sample of low-educated single women (those with a high school degree or less), and using weekly employment as the extensive

6

5

4

-10

-5

-4

-3 -2 -1 0 1 Event Time

2 3 4 5 6

-10

-5 -4 -3 -2 -1

margin measure.

0 1 Event Time

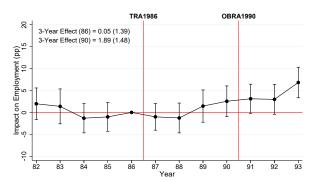
2 3

FIGURE A.25: DID EVENT STUDIES OF FEDERAL EITC REFORMS

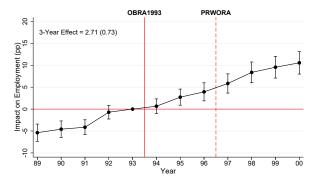
LOW-EDUCATED SINGLE WOMEN, WEEKLY EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT

A: 1975 Reform TRA1975 20 3-Year Effect = 0.73 (1.99) Impact on Employment (pp) 5 0 5 10 15 -10 75 Year 70 72 73 74 76 77 78 79 80 81 71

B: 1986 & 1990 Reforms

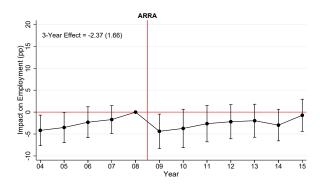


C: 1993 Reform

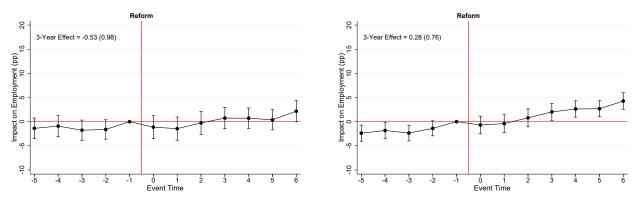


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

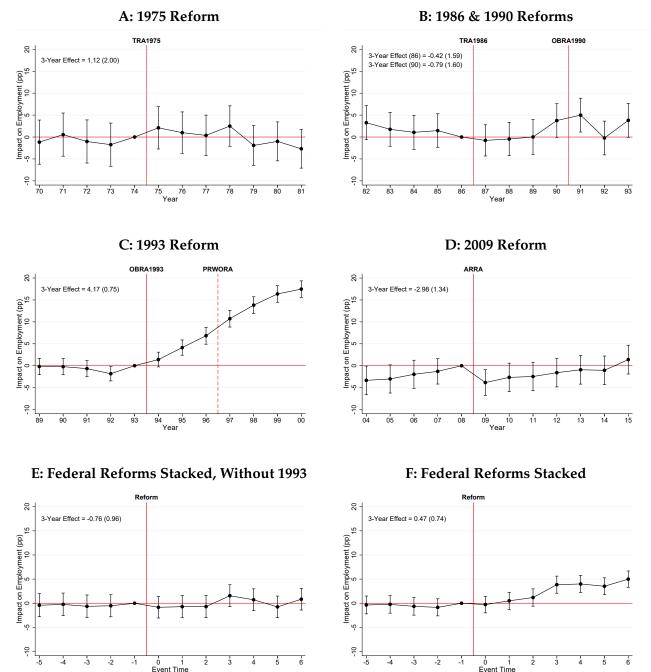






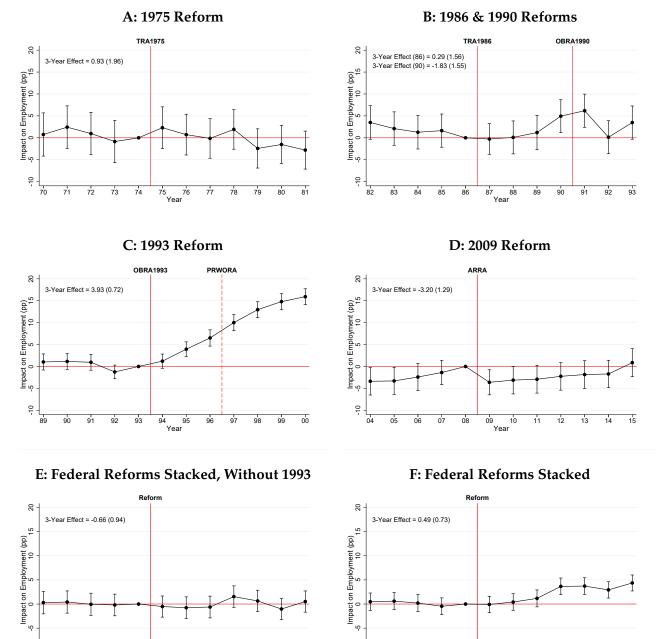
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of low-educated single women (those with a high school degree or less), and using weekly employment as the extensive margin measure.

FIGURE A.26: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT RAW DATA



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the estimation sample. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.27: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT CONTROLS FOR BASIC DEMOGRAPHICS



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the estimation sample. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

6

5

9

-5 -4 -3 -2 -1

0 1 Event Time 2 3

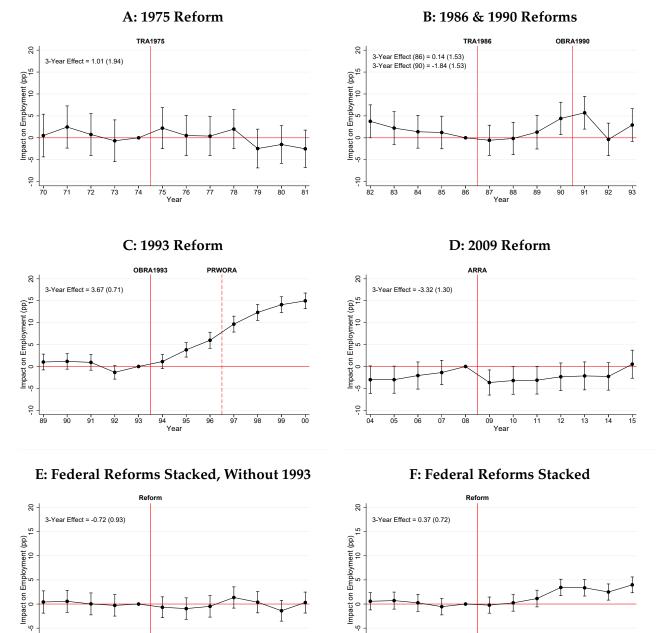
6

-10

-5 -4 -3 -2 -1

0 1 Event Time 2 3

FIGURE A.28: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the estimation sample. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

5 6

9

-5 -4 -3 -2 -1

0 1 Event Time 2

3

6

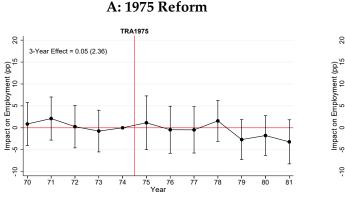
-10

-5 -4 -3 -2 -1

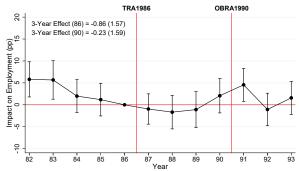
0 1 Event Time 2 3

FIGURE A.29: DID EVENT STUDIES OF FEDERAL EITC REFORMS

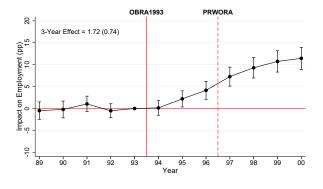
SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

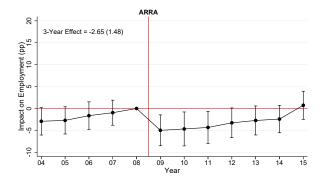


C: 1993 Reform

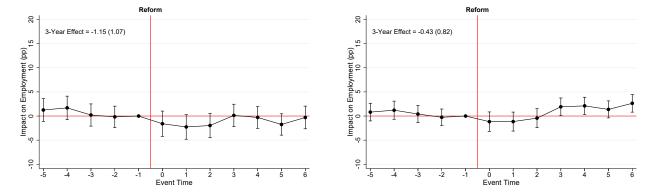


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



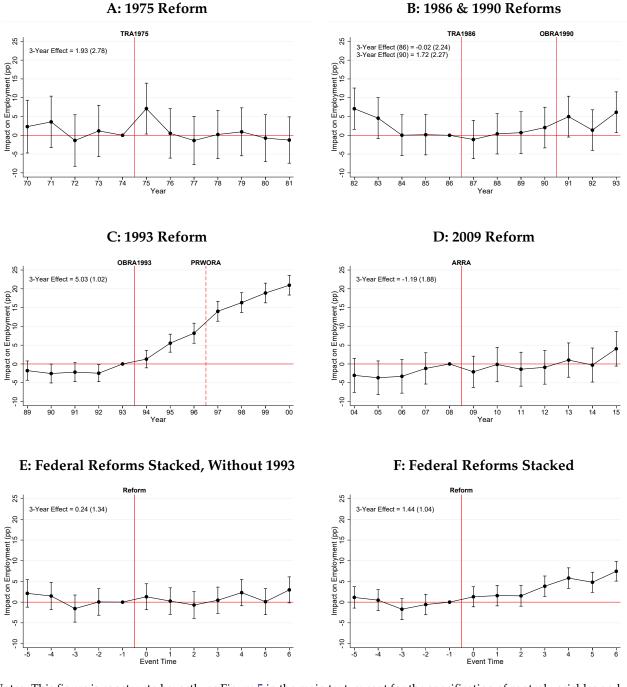
F: Federal Reforms Stacked



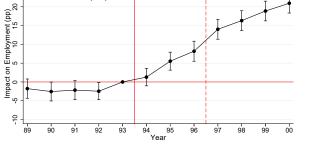
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1). This figure also corresponds to Figure 6 in the main text.

FIGURE A.30: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT RAW DATA

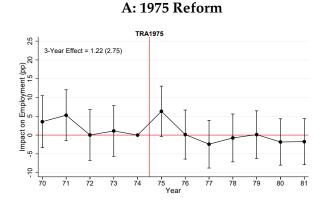


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the estimation sample. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

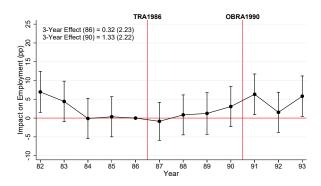


102

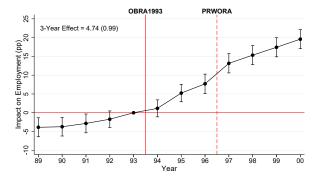
FIGURE A.31: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT CONTROLS FOR BASIC DEMOGRAPHICS



B: 1986 & 1990 Reforms

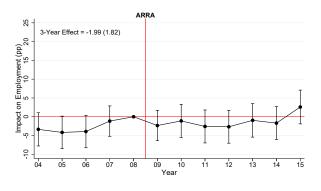


C: 1993 Reform

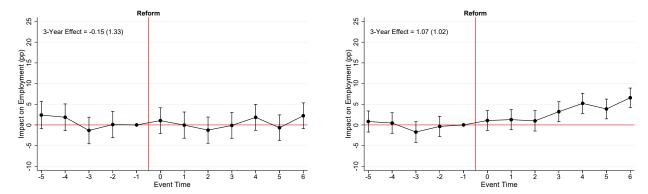


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

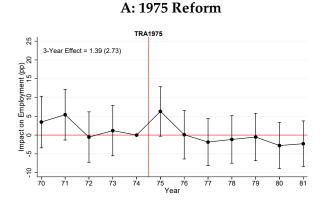


F: Federal Reforms Stacked

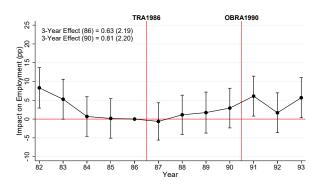


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the estimation sample. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

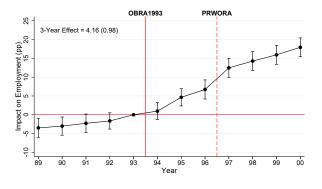
FIGURE A.32: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS



B: 1986 & 1990 Reforms

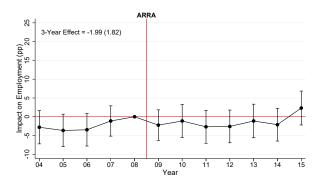


C: 1993 Reform

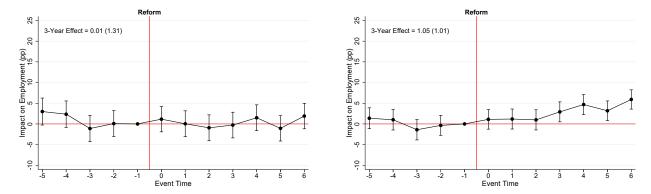


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



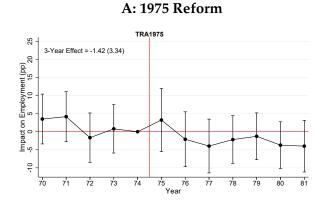
F: Federal Reforms Stacked



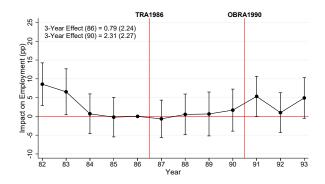
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the estimation sample. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.33: DID EVENT STUDIES OF FEDERAL EITC REFORMS

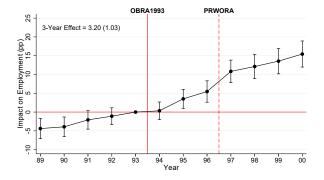
SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

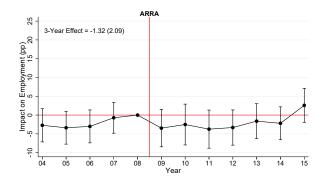


C: 1993 Reform

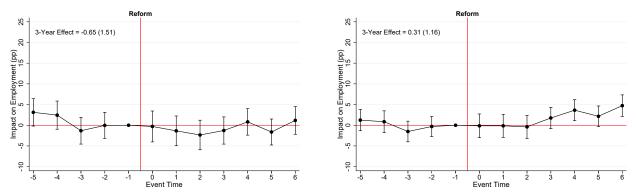


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

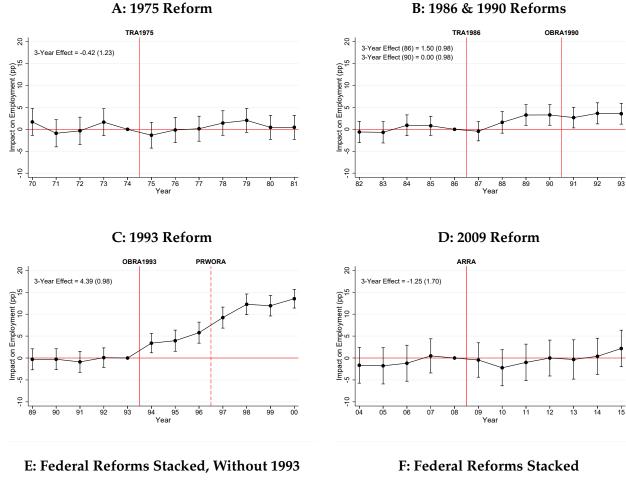


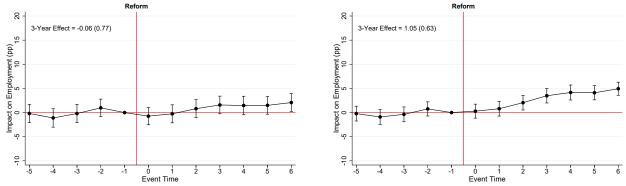
F: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using weekly employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.34: DID EVENT STUDIES OF FEDERAL EITC REFORMS All Single Women, Annual Employment Raw Data

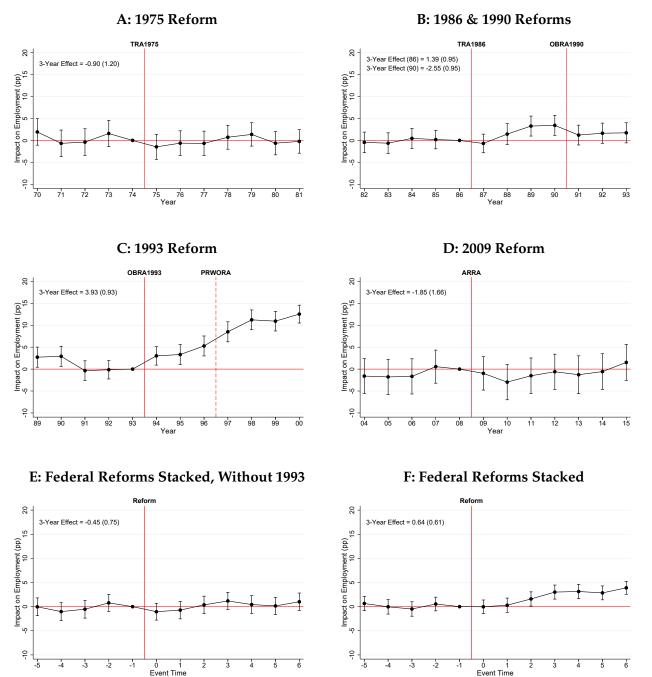




Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) without any controls, the sample of all single women, and using annual employment as the extensive margin measure.

FIGURE A.35: DID EVENT STUDIES OF FEDERAL EITC REFORMS

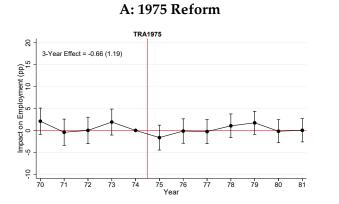
ALL SINGLE WOMEN, ANNUAL EMPLOYMENT CONTROLS FOR BASIC DEMOGRAPHICS



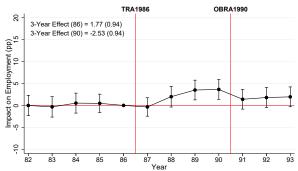
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, the sample of all single women, and using annual employment as the extensive margin measure.

FIGURE A.36: DID EVENT STUDIES OF FEDERAL EITC REFORMS

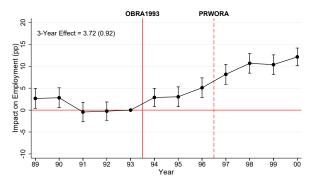
ALL SINGLE WOMEN, ANNUAL EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS



B: 1986 & 1990 Reforms

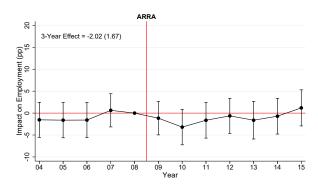


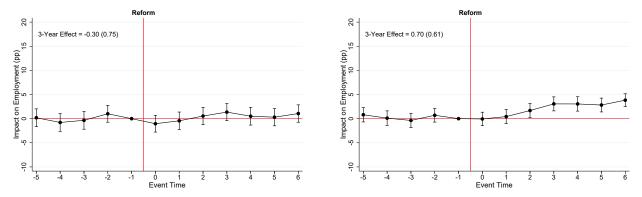
C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



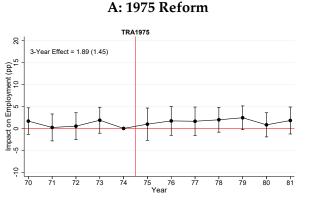


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, the sample of all single women, and using annual employment as the extensive margin measure.

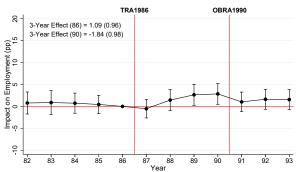
FIGURE A.37: DID EVENT STUDIES OF FEDERAL EITC REFORMS

All Single Women, Annual Employment

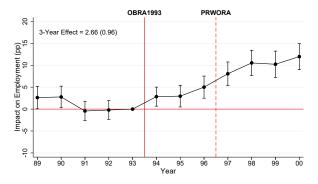
CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

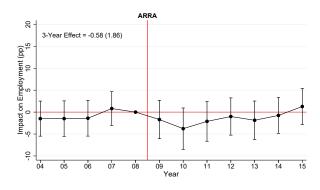


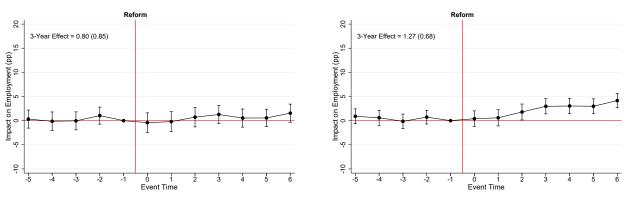
C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

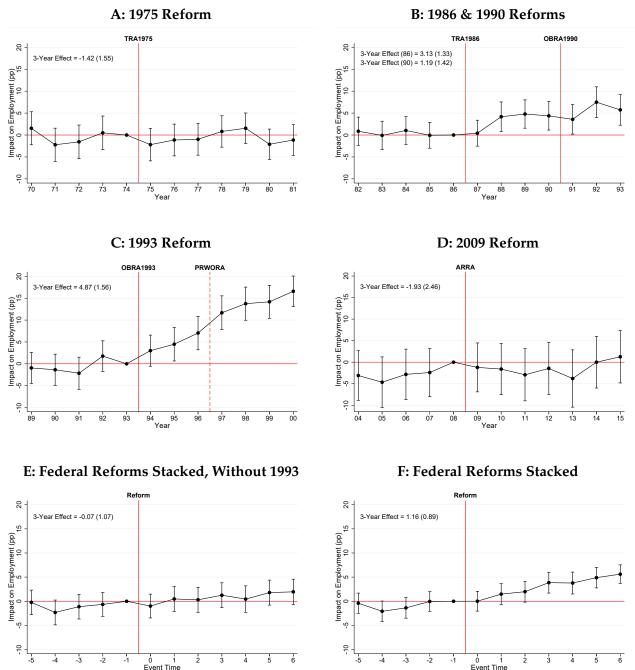
D: 2009 Reform





Notes: This figure is constructed exactly as Figure 5 in the main text, except for the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, the sample of all single women, and using annual employment as the extensive margin measure.





Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of low-educated single women (those with a high school degree or less), and using annual employment as the extensive margin measure.

-5 -4 -3 -2 -1 2 3 4 5

-5 -4 -3 -2 -1 0 1 Event Time

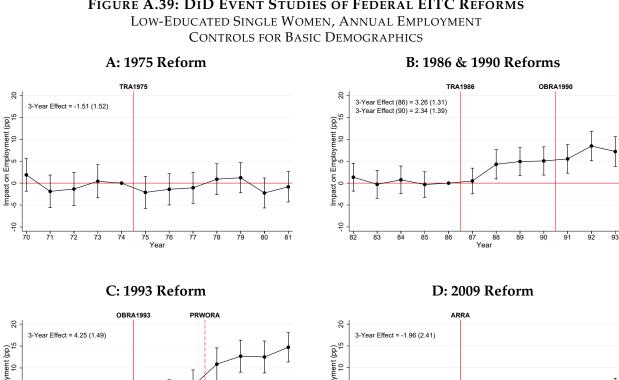
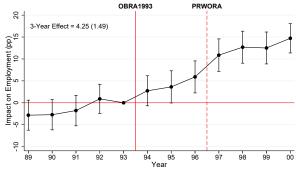
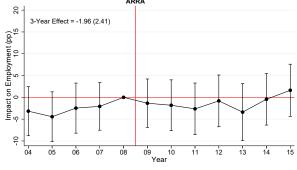


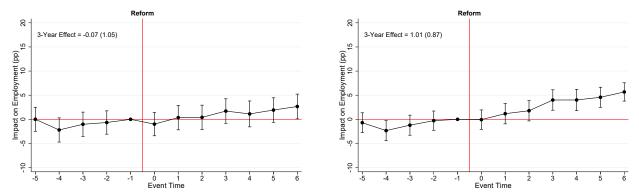
FIGURE A.39: DID EVENT STUDIES OF FEDERAL EITC REFORMS



E: Federal Reforms Stacked, Without 1993



F: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, a sample of low-educated single women (those with a high school degree or less), and using annual employment as the extensive margin measure.

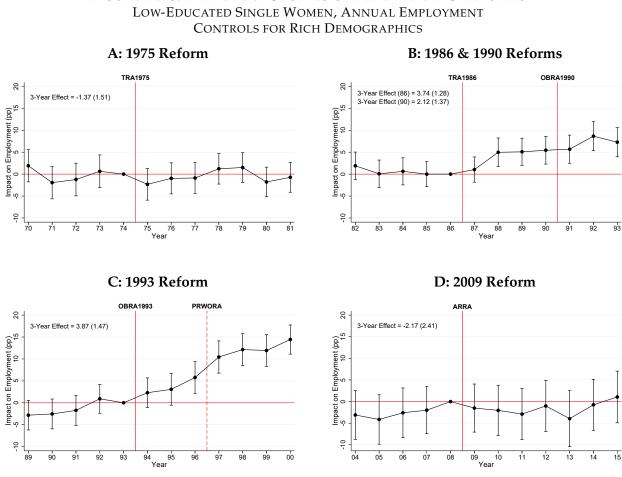


FIGURE A.40: DID EVENT STUDIES OF FEDERAL EITC REFORMS

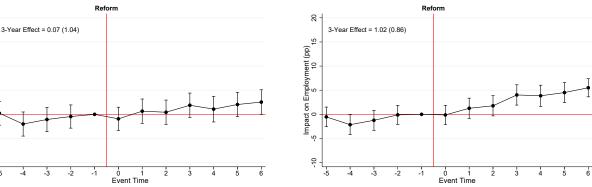
E: Federal Reforms Stacked, Without 1993

20

Impact on Employment (pp) 0 5 10 15

-10

-5 -4



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of low-educated single women (those with a high school degree or less), and using annual employment as the extensive margin measure.

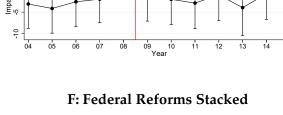
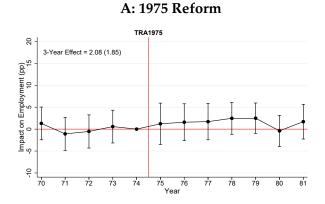
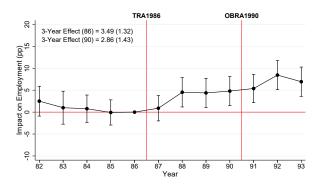


FIGURE A.41: DID EVENT STUDIES OF FEDERAL EITC REFORMS

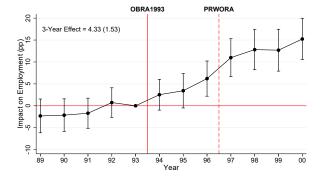
LOW-EDUCATED SINGLE WOMEN, ANNUAL EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

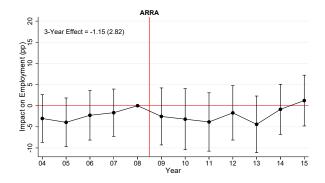


C: 1993 Reform

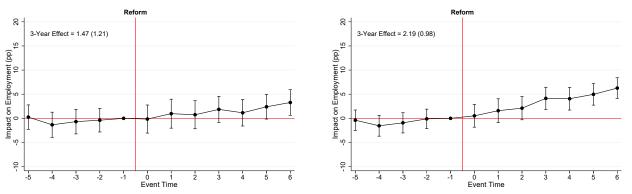


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

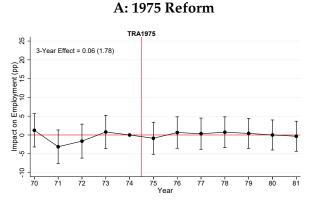


F: Federal Reforms Stacked

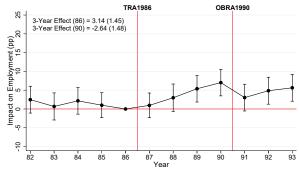


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of low-educated single women (those with a high school degree or less), and using annual employment as the extensive margin measure.

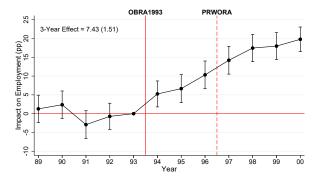
FIGURE A.42: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT RAW DATA



B: 1986 & 1990 Reforms

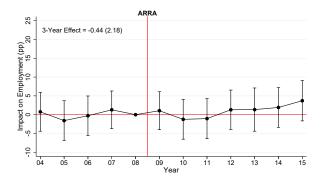


C: 1993 Reform

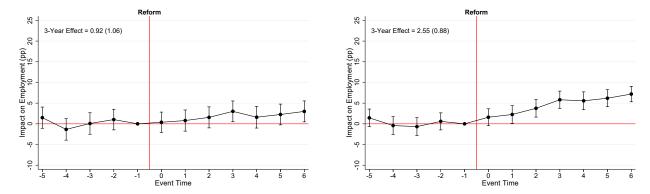


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

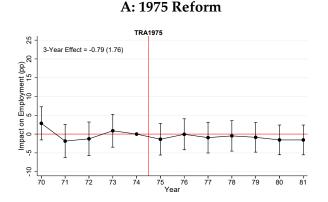


F: Federal Reforms Stacked

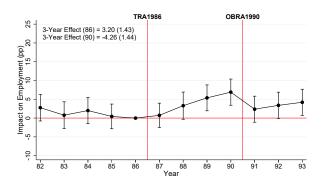


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

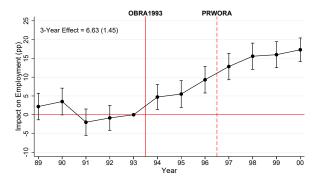
FIGURE A.43: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT CONTROLS FOR BASIC DEMOGRAPHICS



B: 1986 & 1990 Reforms

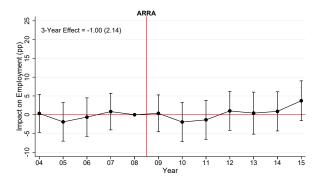


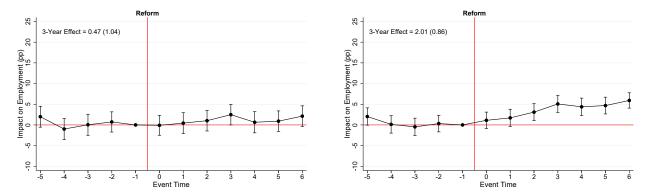
C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

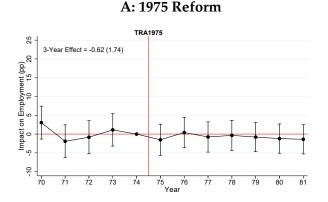
D: 2009 Reform



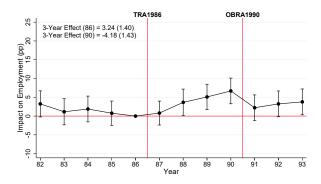


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

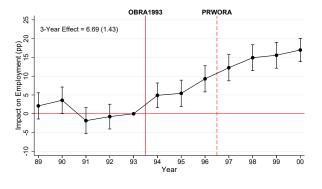
FIGURE A.44: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS



B: 1986 & 1990 Reforms

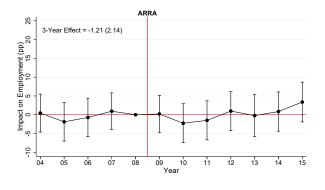


C: 1993 Reform

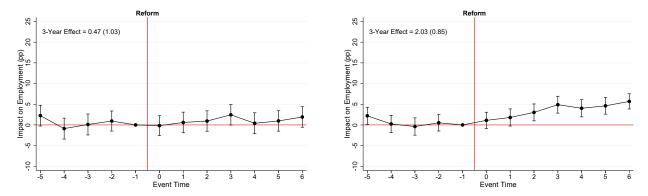


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



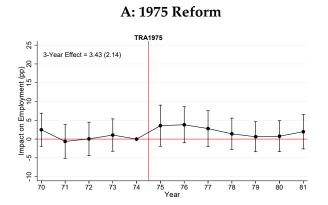
F: Federal Reforms Stacked



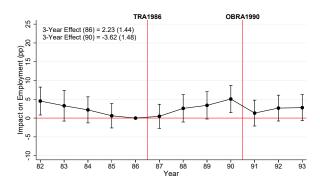
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.45: DID EVENT STUDIES OF FEDERAL EITC REFORMS

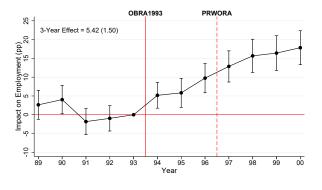
SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

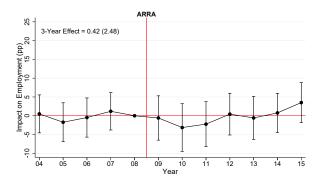


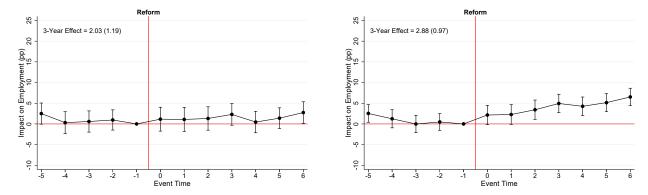
C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

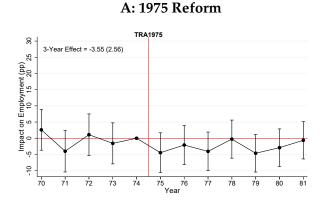




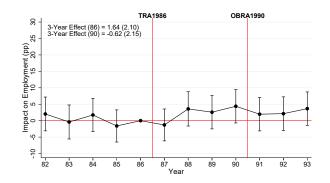
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample and the extensive margin measure. The figure is based on specification (2) with controls for rich demographic and unemployment, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.46: DID EVENT STUDIES OF FEDERAL EITC REFORMS

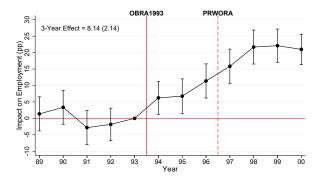
SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT RAW DATA



B: 1986 & 1990 Reforms

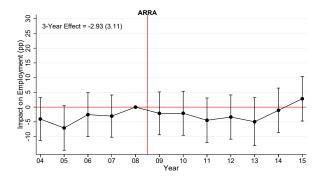


C: 1993 Reform

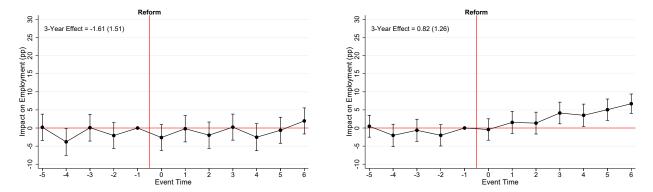


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

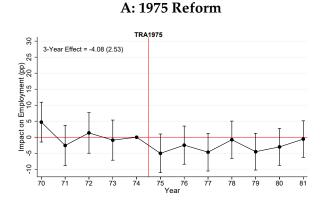


F: Federal Reforms Stacked

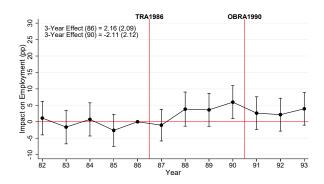


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

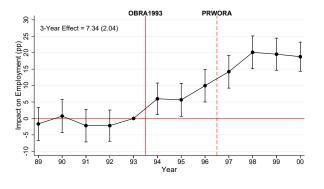
FIGURE A.47: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT CONTROLS FOR BASIC DEMOGRAPHICS



B: 1986 & 1990 Reforms

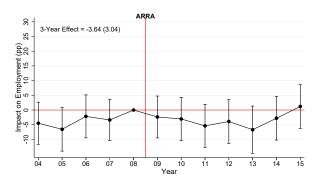


C: 1993 Reform

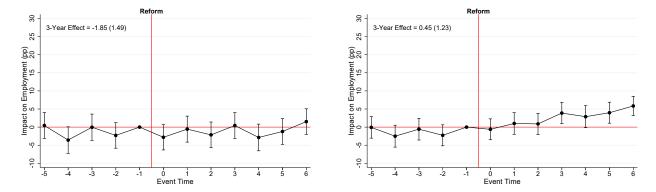


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

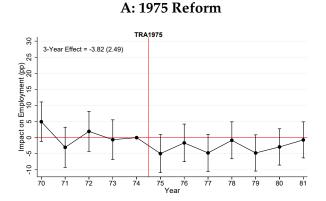


F: Federal Reforms Stacked

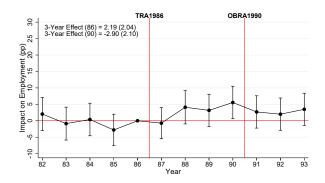


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

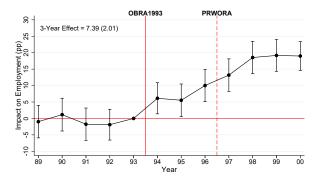
FIGURE A.48: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS



B: 1986 & 1990 Reforms

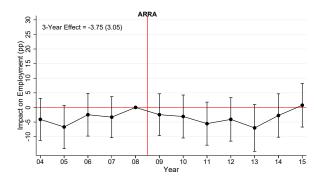


C: 1993 Reform

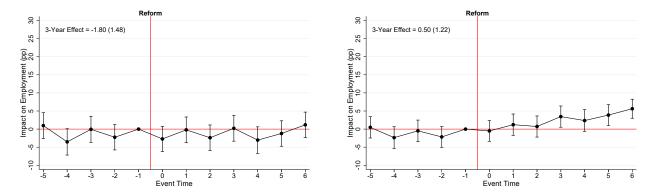


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



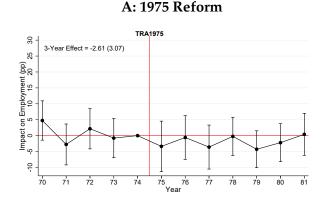
F: Federal Reforms Stacked



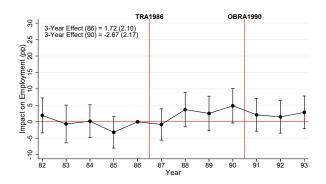
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.49: DID EVENT STUDIES OF FEDERAL EITC REFORMS

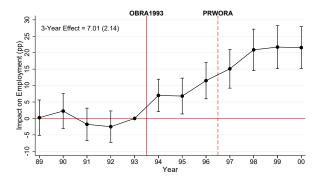
SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL EMPLOYMENT CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

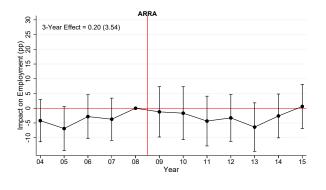


C: 1993 Reform

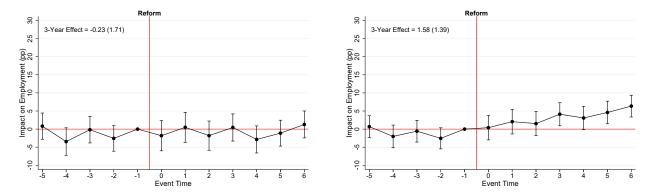


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

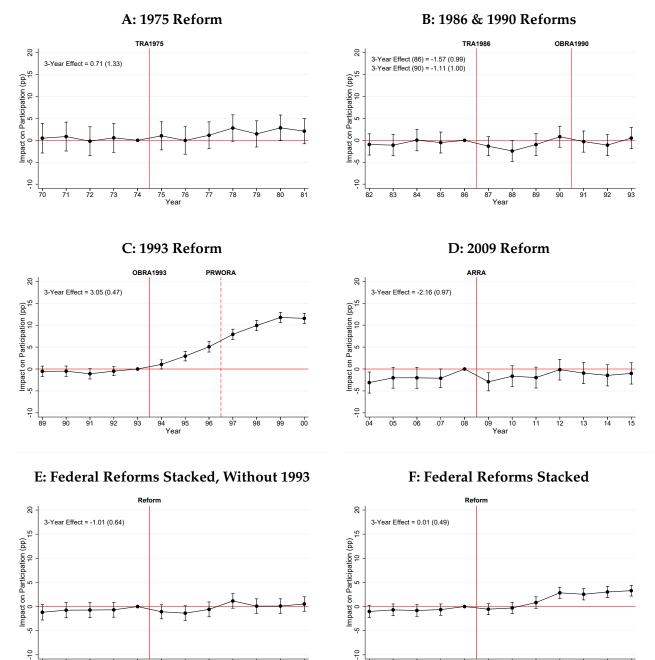


F: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using annual employment as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.50: DID EVENT STUDIES OF FEDERAL EITC REFORMS All Single Women, Weekly Participation Raw Data



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) without any controls, the sample of all single women, and using weekly participation as the extensive margin measure.

122

-5 -4 -3 -2 -1

0 1 Event Time 2 3

6

5

6

5

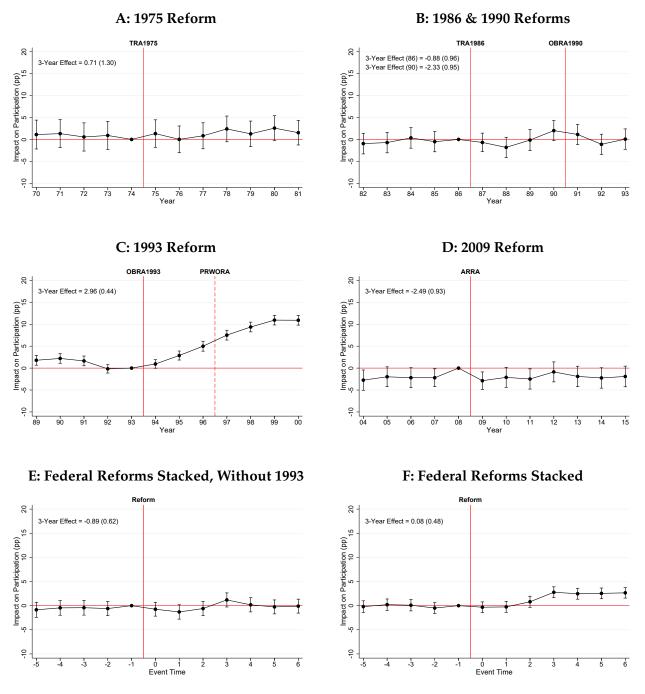
-5

-3 -2 -1

0 1 Event Time

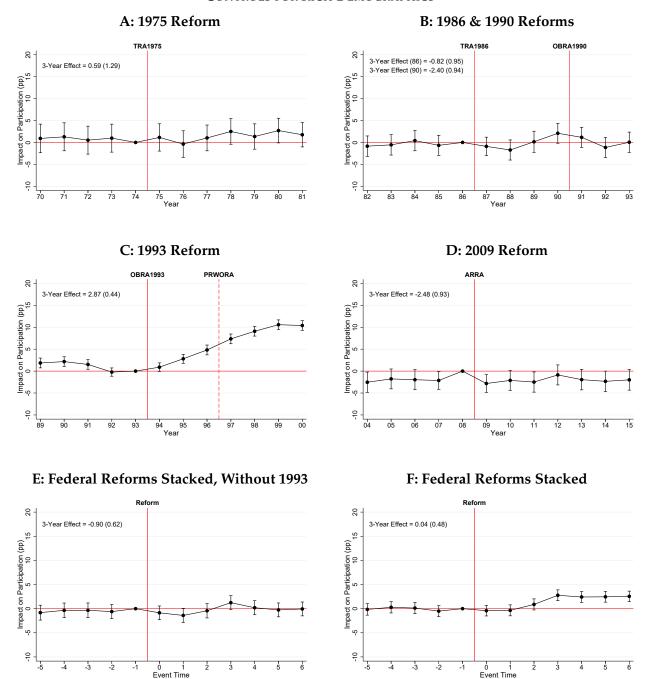
FIGURE A.51: DID EVENT STUDIES OF FEDERAL EITC REFORMS

ALL SINGLE WOMEN, WEEKLY PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, the sample of all single women, and using weekly participation as the extensive margin measure.

FIGURE A.52: DID EVENT STUDIES OF FEDERAL EITC REFORMS All Single Women, Weekly Participation Controls for Rich Demographics

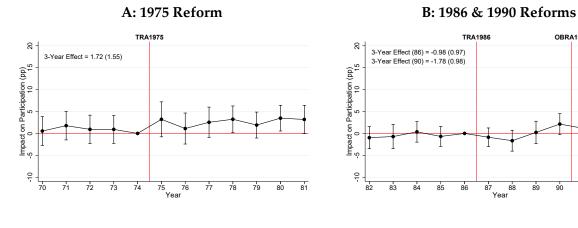


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, the sample of all single women, and using weekly participation as the extensive margin measure.

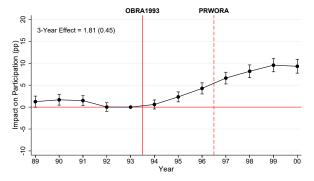
FIGURE A.53: DID EVENT STUDIES OF FEDERAL EITC REFORMS

ALL SINGLE WOMEN, WEEKLY PARTICIPATION

CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

88 89

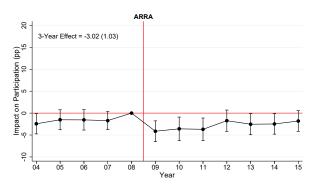
Year

87

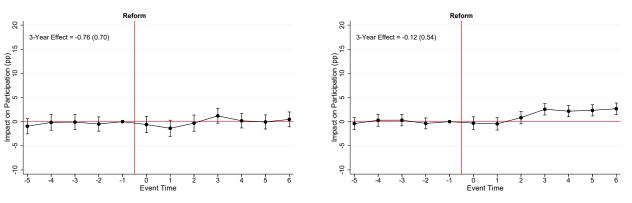
86

TRA1986

OBRA1990

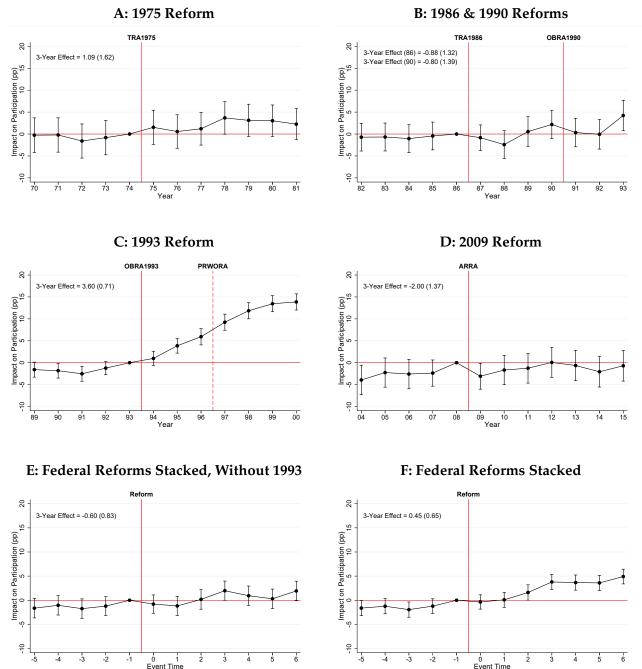






Notes: This figure is constructed exactly as Figure 5 in the main text, except for the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, the sample of all single women, and using weekly participation as the extensive margin measure.





Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, the sample of low-educated single women (those with a high school degree or less), and using weekly participation as the extensive margin measure.

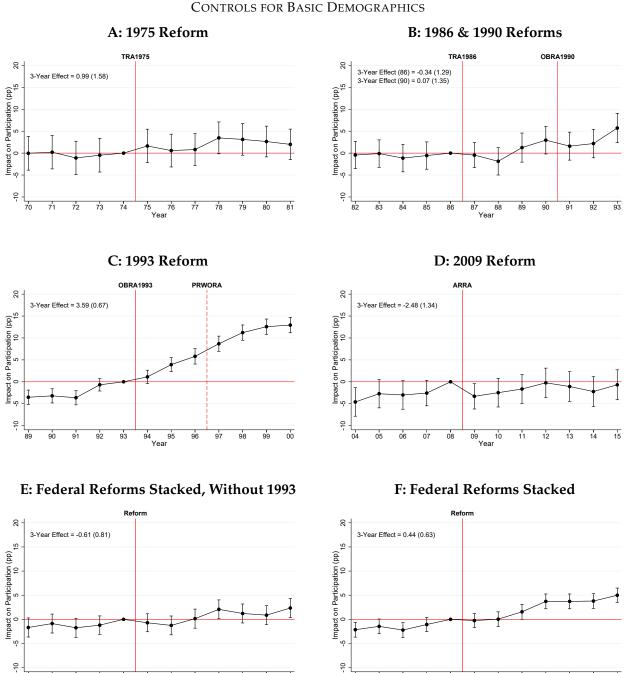


FIGURE A.55: DID EVENT STUDIES OF FEDERAL EITC REFORMS LOW-EDUCATED SINGLE WOMEN, WEEKLY PARTICIPATION

CONTROLS FOR BASIC DEMOGRAPHICS

Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, the sample of low-educated single women (those with a high school degree or less), and using weekly participation as the extensive margin measure.

-5 -4 -3 -2 0 1 Event Time

-1

2 3 4 5 6

6

-5 -4 -3 -2 -1 0 1 Event Time

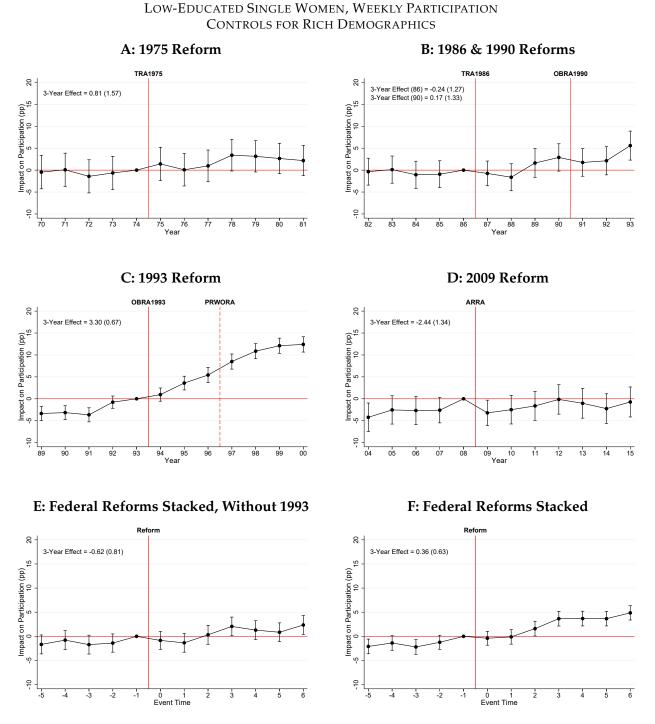
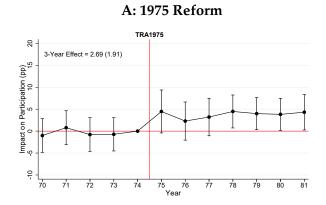


FIGURE A.56: DID EVENT STUDIES OF FEDERAL EITC REFORMS

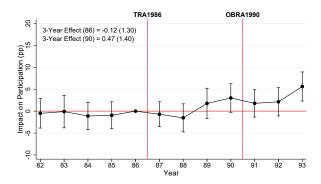
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, the sample of low-educated single women (those with a high school degree or less), and using weekly participation as the extensive margin measure.

FIGURE A.57: DID EVENT STUDIES OF FEDERAL EITC REFORMS

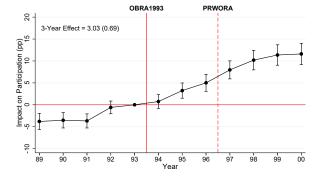
LOW-EDUCATED SINGLE WOMEN, WEEKLY PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

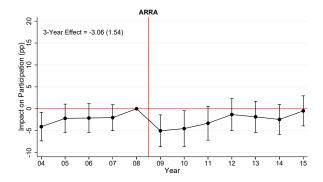


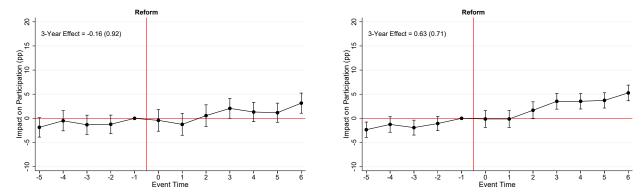
C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

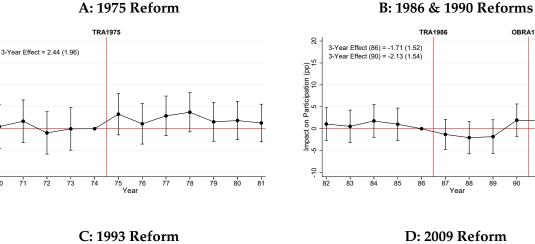
D: 2009 Reform





Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, the sample of low-educated single women (those with a high school degree or less), and using weekly participation as the extensive margin measure.

FIGURE A.58: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY PARTICIPATION RAW DATA



OBRA1993 PRWORA 20 3-Year Effect = 4.10 (0.71) Impact on Participation (pp) 5 0 5 10 15 ŝ -10 94 Year 00 89 90 91 92 93 95 96 97 98 99

20

(pp) 15

Impact on Participation (5 0 5 10

9

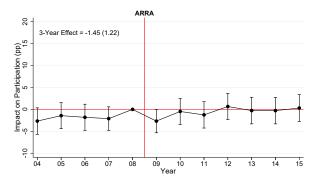
70

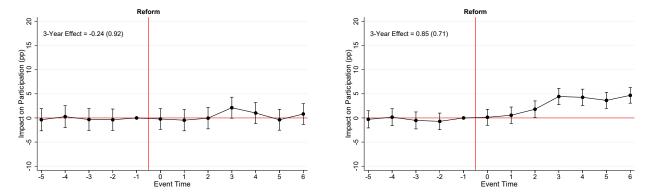
E: Federal Reforms Stacked, Without 1993

D: 2009 Reform

OBRA1990

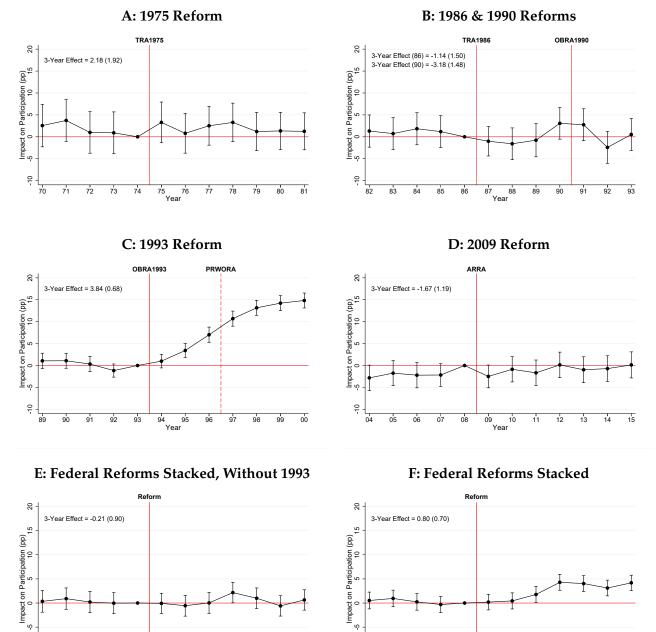
90 91 92 93





Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.59: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

6

5

9

-5 -4 -3 -2 -1

0 1 Event Time 2 3

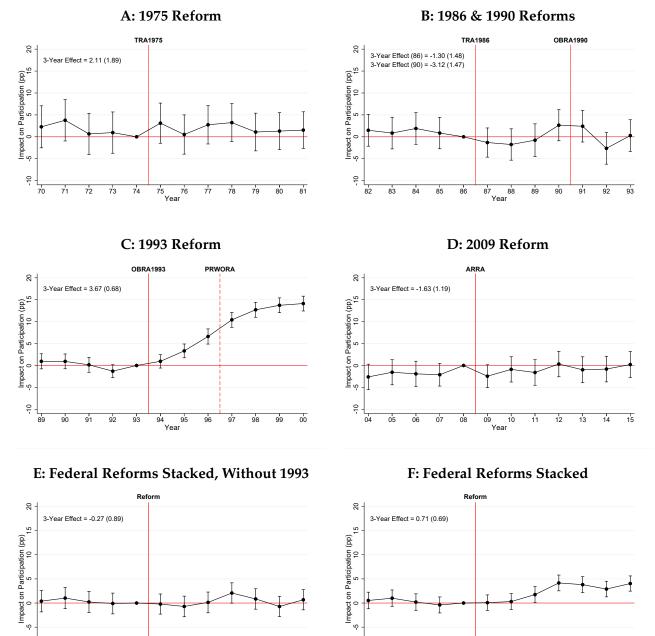
6

-10

-5 -4 -3 -2 -1

0 1 Event Time

FIGURE A.60: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

6

5

9

-5 -4 -3 -2 -1

0 1 Event Time 2 3

6

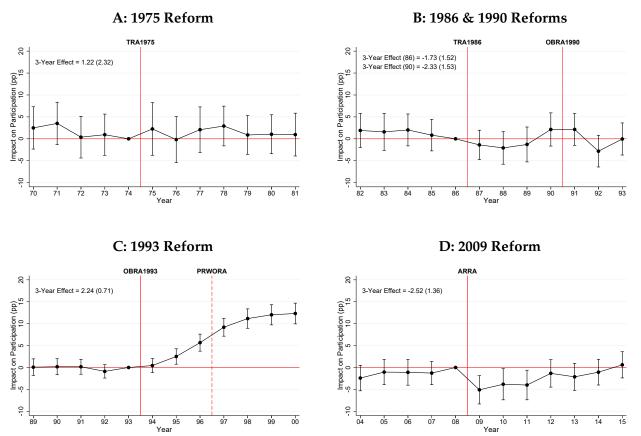
-10

-5 -4 -3 -2 -1

0 1 Event Time

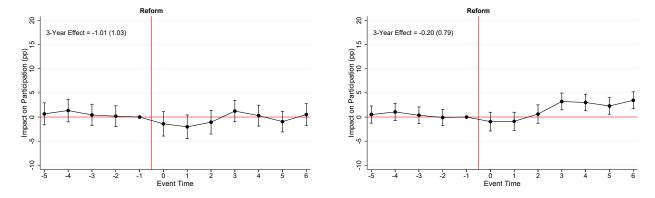
FIGURE A.61: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



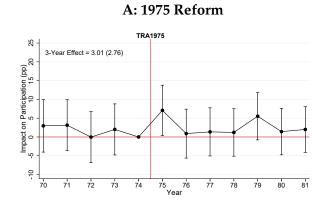
E: Federal Reforms Stacked, Without 1993

F: Federal Reforms Stacked

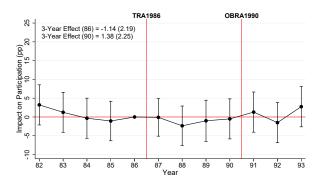


Notes: This figure is constructed exactly as Figure 5 in the main text, except for estimation sample and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

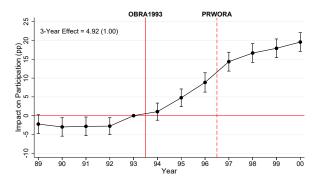
FIGURE A.62: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY PARTICIPATION RAW DATA



B: 1986 & 1990 Reforms

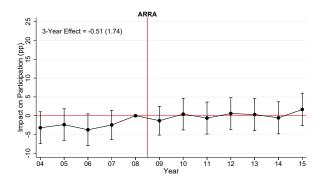


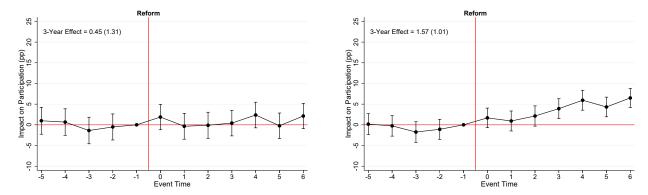
C: 1993 Reform



E: Federal Reforms Stacked, Without 1993

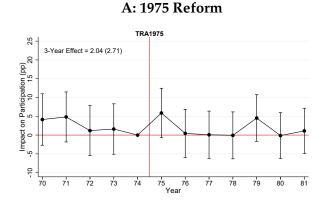
D: 2009 Reform



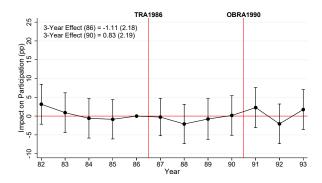


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

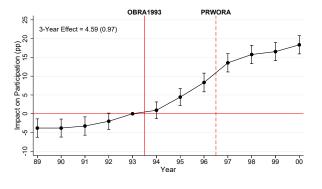
FIGURE A.63: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS



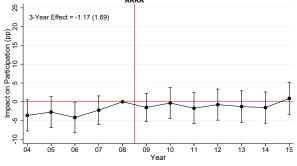
B: 1986 & 1990 Reforms



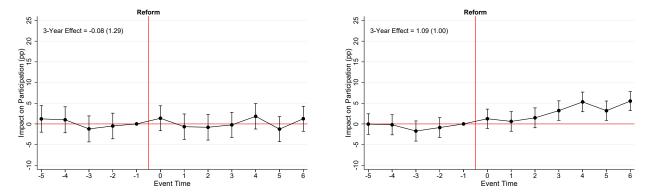
C: 1993 Reform



D: 2009 Reform

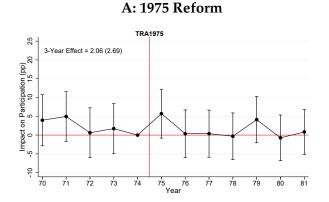


E: Federal Reforms Stacked, Without 1993

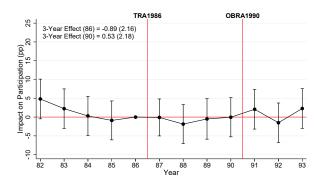


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

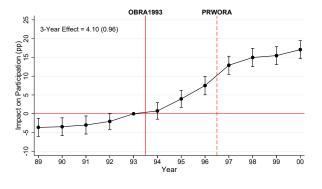
FIGURE A.64: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS



B: 1986 & 1990 Reforms

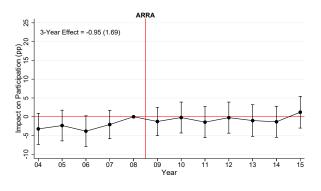


C: 1993 Reform

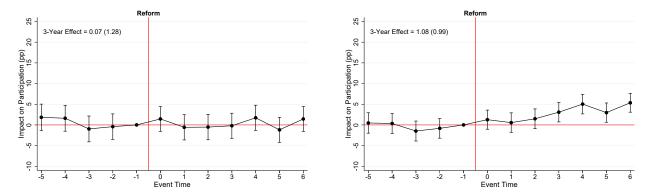


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



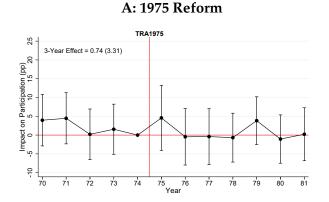
F: Federal Reforms Stacked



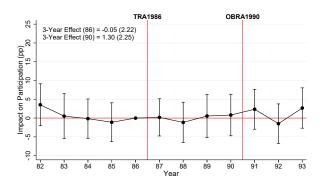
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.65: DID EVENT STUDIES OF FEDERAL EITC REFORMS

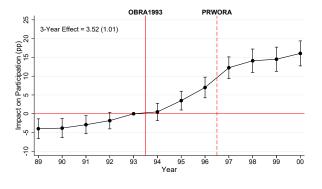
SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, WEEKLY PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



B: 1986 & 1990 Reforms

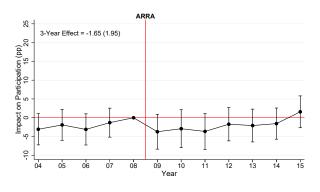


C: 1993 Reform

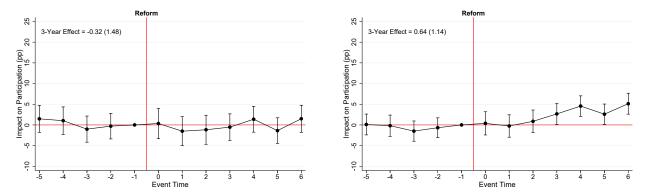


E: Federal Reforms Stacked, Without 1993

D: 2009 Reform



F: Federal Reforms Stacked

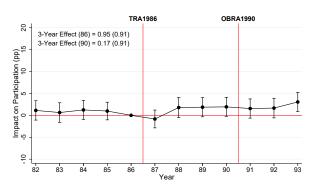


Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using weekly participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.66: DID EVENT STUDIES OF FEDERAL EITC REFORMS

ALL SINGLE WOMEN, ANNUAL PARTICIPATION

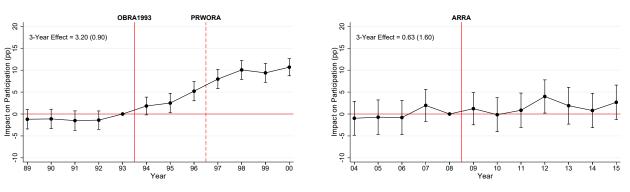
RAW DATA



A: 1986 & 1990 Reforms

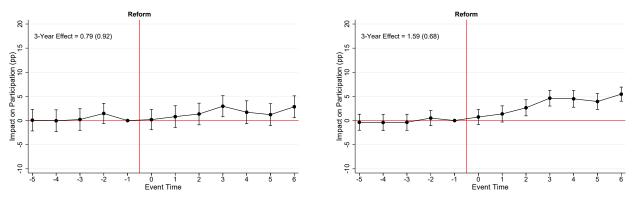
B: 1993 Reform

C: 2009 Reform



D: Federal Reforms Stacked, Without 1993

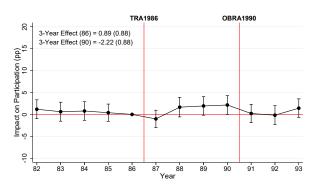
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) without any controls, the sample of all single women, and using annual participation as the extensive margin measure.

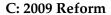
FIGURE A.67: DID EVENT STUDIES OF FEDERAL EITC REFORMS

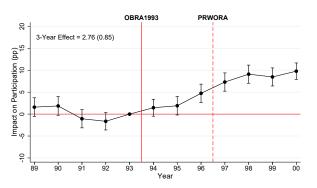
ALL SINGLE WOMEN, ANNUAL PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS



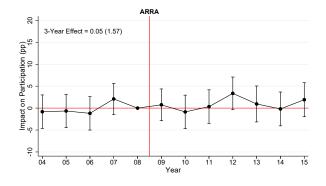
A: 1986 & 1990 Reforms

B: 1993 Reform

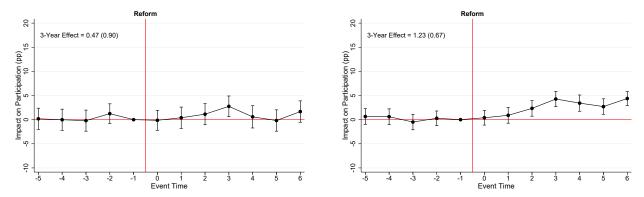




D: Federal Reforms Stacked, Without 1993



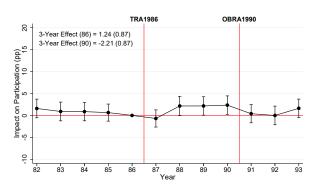
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, the sample of all single women, and using annual participation as the extensive margin measure.

FIGURE A.68: DID EVENT STUDIES OF FEDERAL EITC REFORMS

ALL SINGLE WOMEN, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS



A: 1986 & 1990 Reforms

B: 1993 Reform

PRWORA

OBRA1993

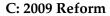
20

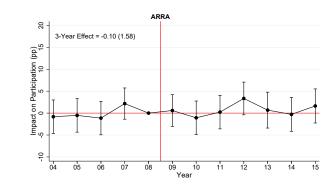
Impact on Participation (pp) 5 0 5 10 15

-10

89 90 91 92 93

3-Year Effect = 2.57 (0.85)



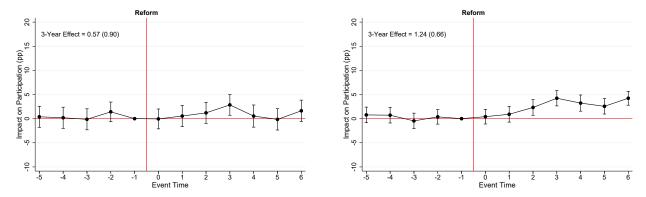


D: Federal Reforms Stacked, Without 1993

94 Year

95 96 97 98 99 00

E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, the sample of all single women, and using annual participation as the extensive margin measure.

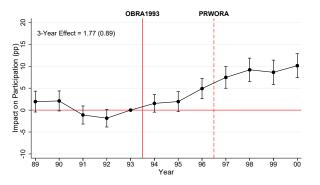
FIGURE A.69: DID EVENT STUDIES OF FEDERAL EITC REFORMS

ALL SINGLE WOMEN, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT

OBRA1990 TRA1986 20 3-Year Effect (86) = 1.05 (0.89) 3-Year Effect (90) = -1.78 (0.91) Impact on Participation (pp) 5 0 5 10 15 -10 87 Year 82 84 86 88 89 90 91 92 93 83 85

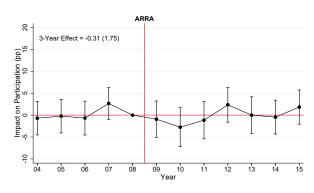
A: 1986 & 1990 Reforms

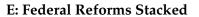


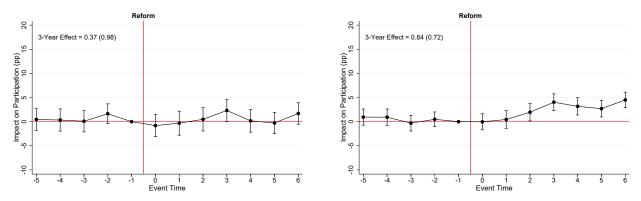


D: Federal Reforms Stacked, Without 1993

C: 2009 Reform



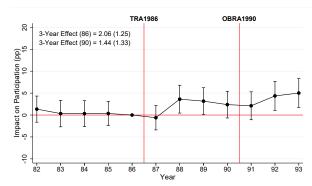




Notes: This figure is constructed exactly as Figure 5 in the main text, except for the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, the sample of all single women, and using annual participation as the extensive margin measure.

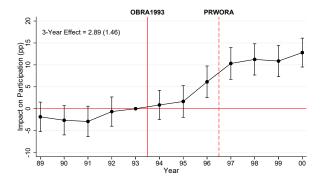
FIGURE A.70: DID EVENT STUDIES OF FEDERAL EITC REFORMS

LOW-EDUCATED SINGLE WOMEN, ANNUAL PARTICIPATION RAW DATA



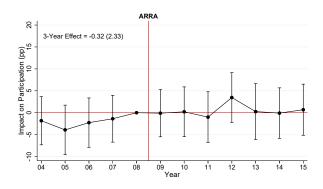
A: 1986 & 1990 Reforms

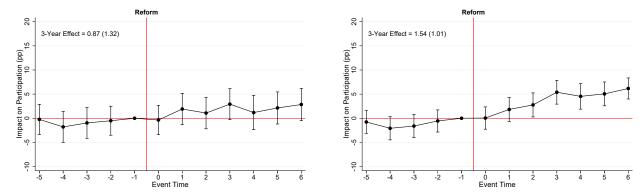
B: 1993 Reform



D: Federal Reforms Stacked, Without 1993

C: 2009 Reform

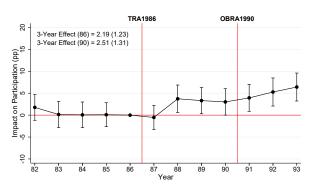




Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, the sample of low-educated single women (those with a high school degree or less), and using annual participation as the extensive margin measure.

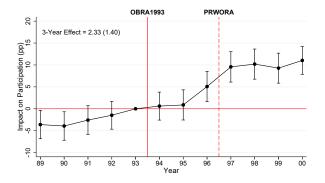
FIGURE A.71: DID EVENT STUDIES OF FEDERAL EITC REFORMS

LOW-EDUCATED SINGLE WOMEN, ANNUAL PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS



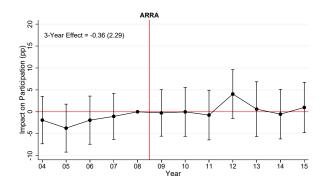
A: 1986 & 1990 Reforms

B: 1993 Reform

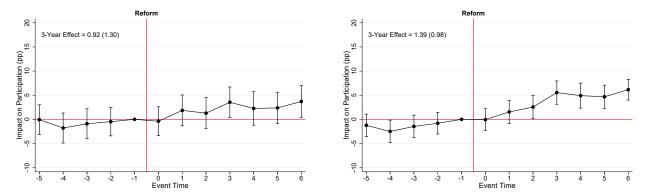


D: Federal Reforms Stacked, Without 1993

C: 2009 Reform



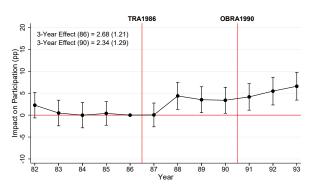
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, the sample of low-educated single women (those with a high school degree or less), and using annual participation as the extensive margin measure.

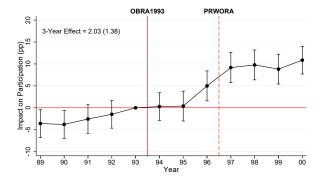
FIGURE A.72: DID EVENT STUDIES OF FEDERAL EITC REFORMS

LOW-EDUCATED SINGLE WOMEN, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS



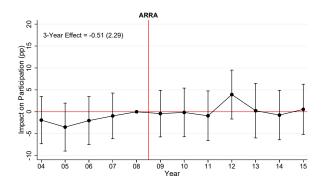
A: 1986 & 1990 Reforms

B: 1993 Reform

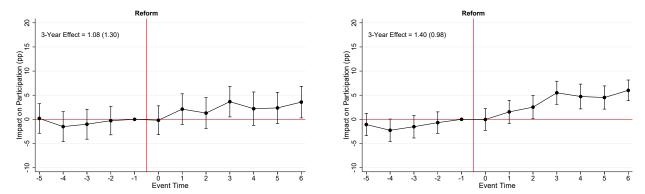


D: Federal Reforms Stacked, Without 1993

C: 2009 Reform



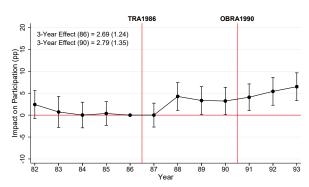
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, the sample of low-educated single women (those with a high school degree or less), and using annual participation as the extensive margin measure.

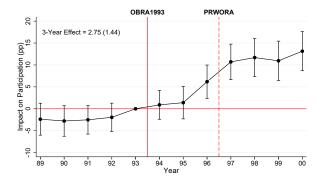
FIGURE A.73: DID EVENT STUDIES OF FEDERAL EITC REFORMS

LOW-EDUCATED SINGLE WOMEN, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



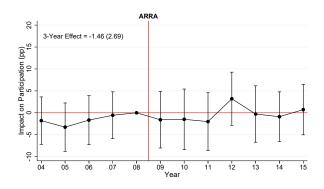
A: 1986 & 1990 Reforms

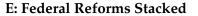
B: 1993 Reform

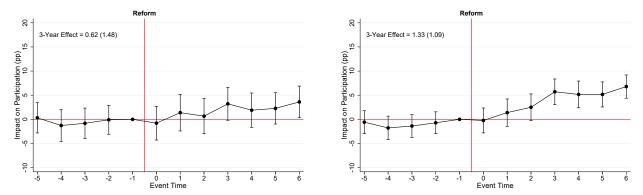


D: Federal Reforms Stacked, Without 1993

C: 2009 Reform



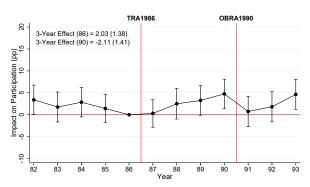




Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, the sample of low-educated single women (those with a high school degree or less), and using annual participation as the extensive margin measure.

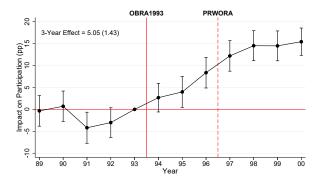
FIGURE A.74: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL PARTICIPATION RAW DATA



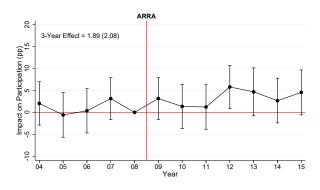
A: 1986 & 1990 Reforms

B: 1993 Reform

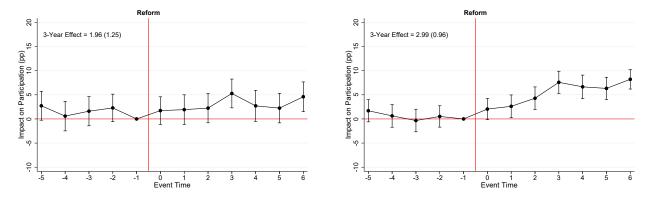


D: Federal Reforms Stacked, Without 1993

C: 2009 Reform



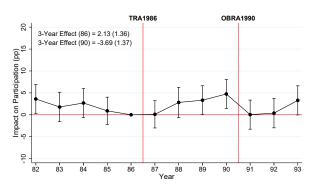
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.75: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS



A: 1986 & 1990 Reforms

B: 1993 Reform

PRWORA

OBRA1993

20

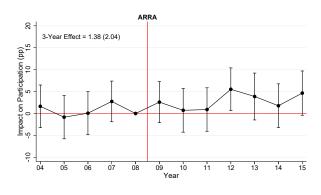
Impact on Participation (pp) 5 0 5 10 15

-10

89 90 91 92 93 94 95 96 97 98 99

3-Year Effect = 4.33 (1.38)

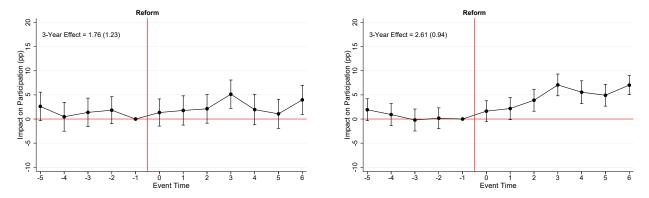




D: Federal Reforms Stacked, Without 1993

Year

E: Federal Reforms Stacked

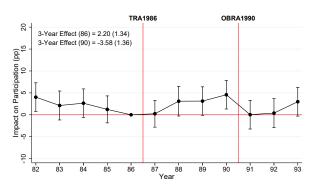


00

Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (5) with controls for basic demographics only, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.76: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS



A: 1986 & 1990 Reforms

B: 1993 Reform

PRWORA

OBRA1993

20

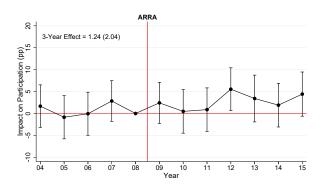
Impact on Participation (pp) 5 0 5 10 15

-10

89 90 91 92 93 94 95 96 97 98 99

3-Year Effect = 4 43 (1 36)

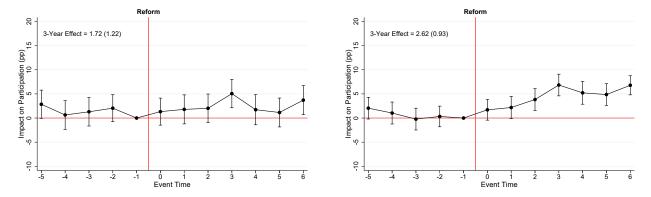






Year

E: Federal Reforms Stacked

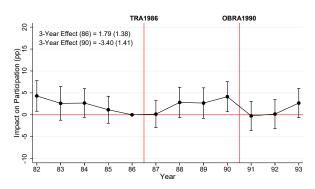


00

Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

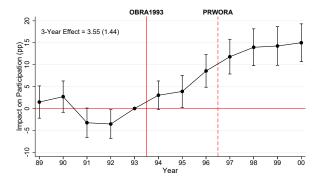
FIGURE A.77: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT



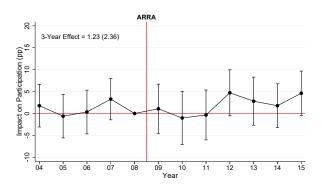
A: 1986 & 1990 Reforms

B: 1993 Reform

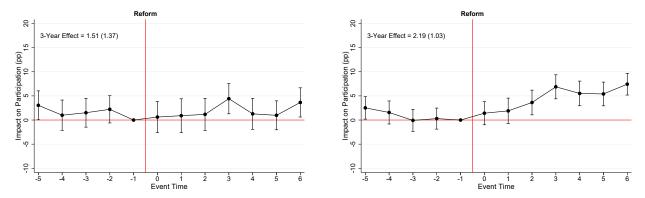


E: Federal Reforms Stacked, Without 1993

C: 2009 Reform



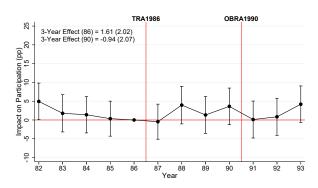
F: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

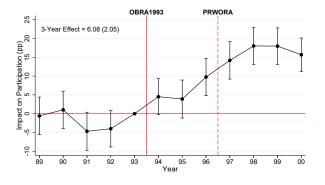
FIGURE A.78: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL PARTICIPATION RAW DATA

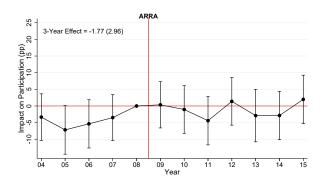


A: 1986 & 1990 Reforms

B: 1993 Reform

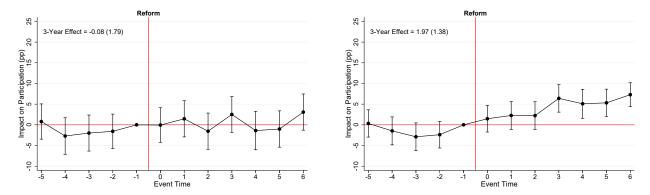


C: 2009 Reform



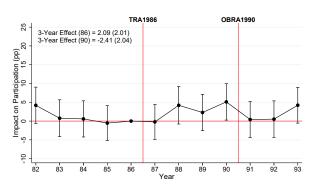
D: Federal Reforms Stacked, Without 1993

E: Federal Reforms Stacked



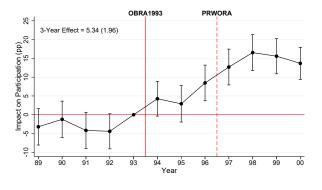
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) without any controls, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.79: DID EVENT STUDIES OF FEDERAL EITC REFORMS SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL PARTICIPATION CONTROLS FOR BASIC DEMOGRAPHICS

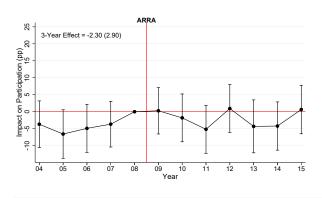


A: 1986 & 1990 Reforms

B: 1993 Reform

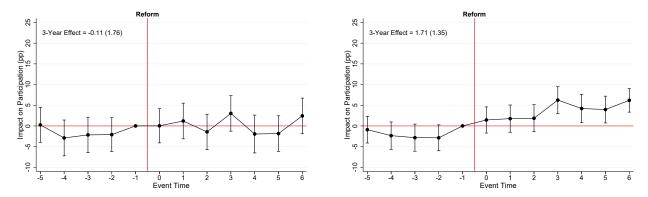


C: 2009 Reform



D: Federal Reforms Stacked, Without 1993

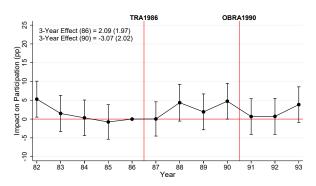
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for basic demographics only, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

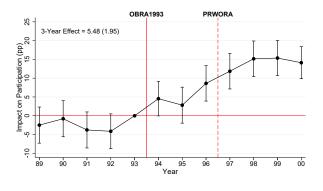
FIGURE A.80: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS

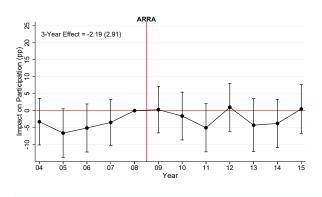


A: 1986 & 1990 Reforms

B: 1993 Reform

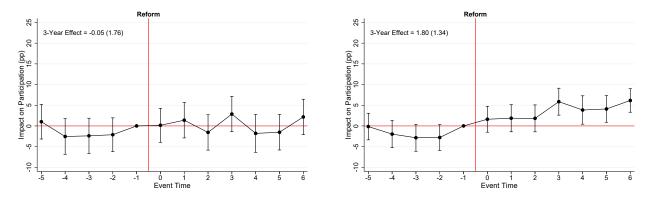


C: 2009 Reform



D: Federal Reforms Stacked, Without 1993

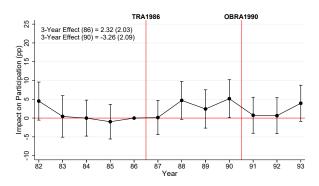
E: Federal Reforms Stacked



Notes: This figure is constructed exactly as Figure 5 in the main text, except for the specification of control variables, estimation sample and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics only, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).

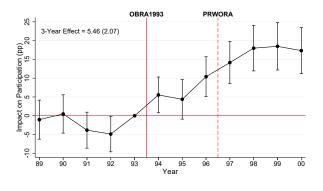
FIGURE A.81: DID EVENT STUDIES OF FEDERAL EITC REFORMS

SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS, ANNUAL PARTICIPATION CONTROLS FOR RICH DEMOGRAPHICS AND UNEMPLOYMENT

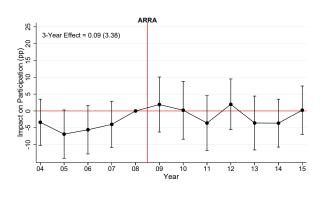


A: 1986 & 1990 Reforms

B: 1993 Reform

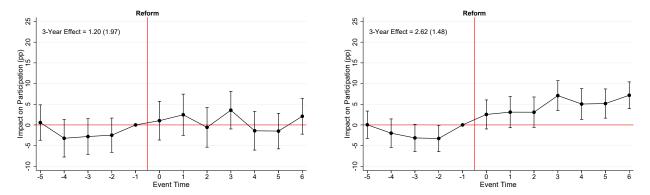


C: 2009 Reform



E: Federal Reforms Stacked, Without 1993

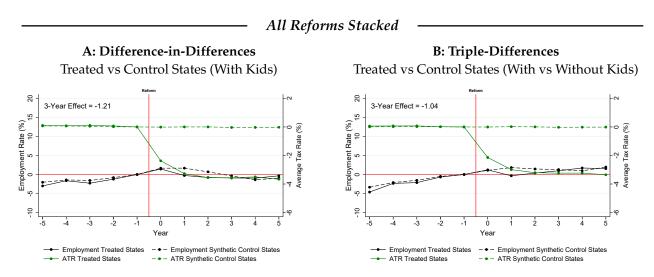
F: Federal Reforms Stacked



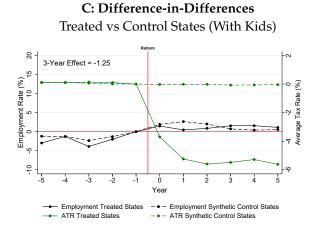
Notes: This figure is constructed exactly as Figure 5 in the main text, except for the estimation sample, and the extensive margin measure. The figure is based on specification (2) with controls for rich demographics and unemployment, a sample of single women with low predicted earnings, and using annual participation as the extensive margin measure. The low-earnings sample includes single women in the bottom quartile of the within-year distribution of predicted earnings, estimated using equation (1).



WEEKLY EMPLOYMENT FOR ALL SINGLE WOMEN



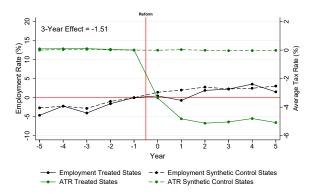
Ten Largest Reforms Stacked



Notes: This figure is exactly the same as Figure 11 in the main text.

D: Triple-Differences

Treated vs Control States (With vs Without Kids)



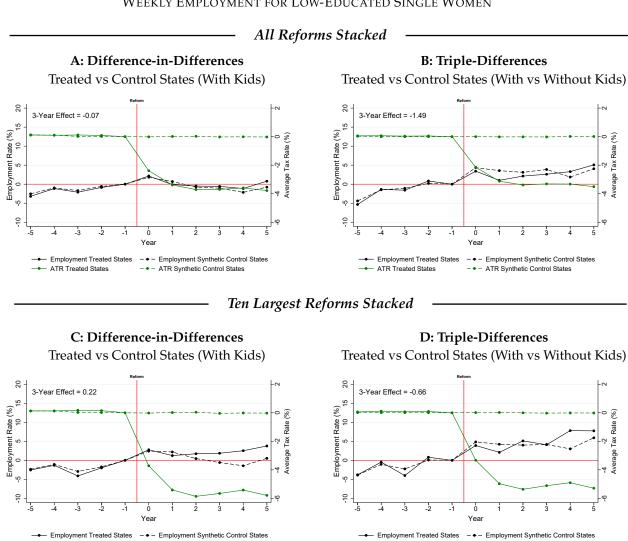


FIGURE A.83: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS WEEKLY EMPLOYMENT FOR LOW-EDUCATED SINGLE WOMEN

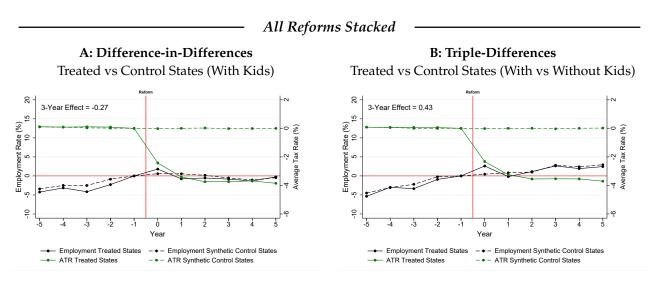
Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample. In this figure, the sample is low-educated single women (those with a high school degree or less) and the extensive margin measure is weekly employment.

ATR Treated States

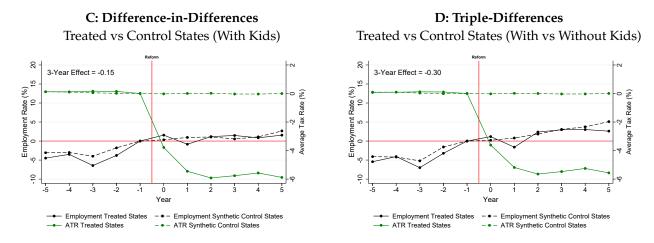
- - ATR Synthetic Control States

- ATR Treated States

FIGURE A.84: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS WEEKLY EMPLOYMENT FOR SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS

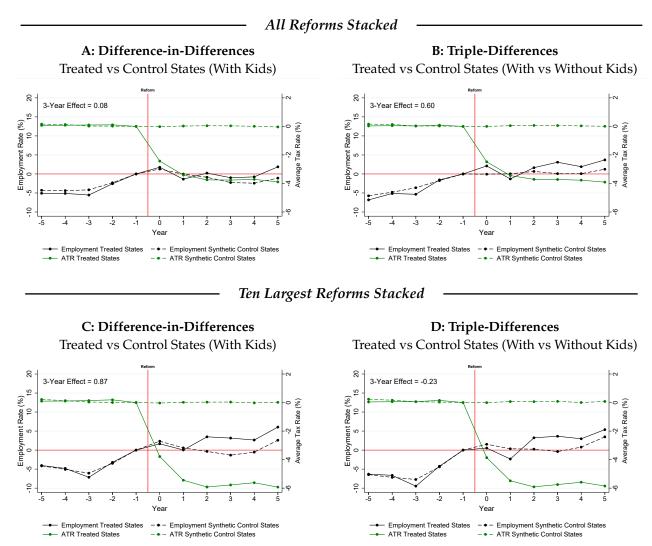


Ten Largest Reforms Stacked



Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is weekly employment. The low-earnings sample includes single women in the bottom half of the within-year and within-state distribution of predicted earnings, estimated using equation (1).





Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is weekly employment. The low-earnings sample includes single women in the bottom quartile of the within-year and within-state distribution of predicted earnings, estimated using equation (1).

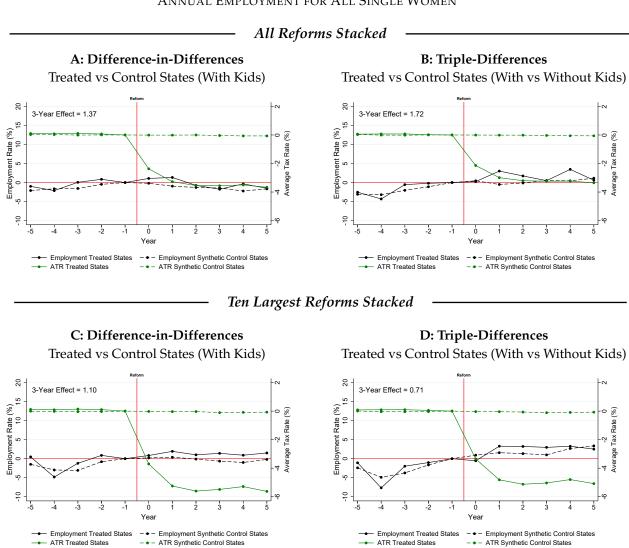
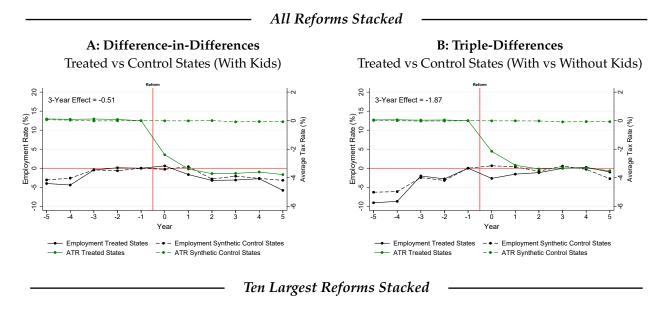
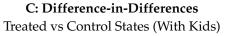


FIGURE A.86: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS ANNUAL EMPLOYMENT FOR ALL SINGLE WOMEN

Notes: This figure is constructed exactly as Figure 11 in the main text, except for the extensive margin measure. In this figure, the sample is all single women and the extensive margin measure is annual employment.

FIGURE A.87: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS ANNUAL EMPLOYMENT FOR LOW-EDUCATED SINGLE WOMEN





ò

Year

- Employment Synthetic Control States

- - ATR Synthetic Control States

20

Employment Rate (%) 5 10 15

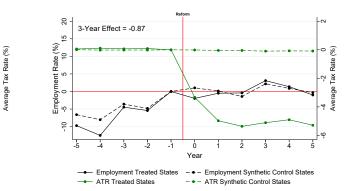
9

3-Year Effect = -1.49

Employment Treated States

ATR Treated States

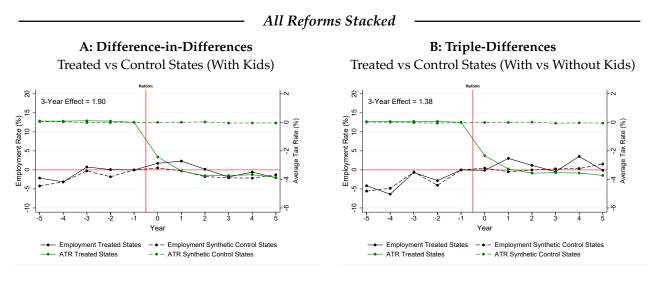
D: Triple-Differences Treated vs Control States (With vs Without Kids)



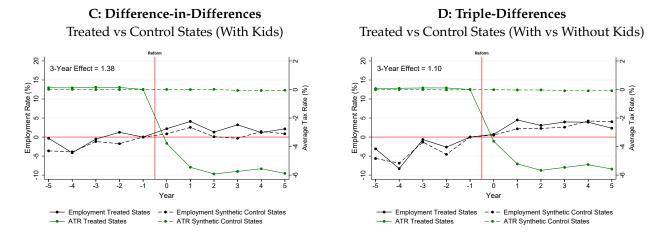
Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is low-educated single women (those with a high school degree or less) and the extensive margin measure is annual employment.

FIGURE A.88: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS

ANNUAL EMPLOYMENT FOR SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS

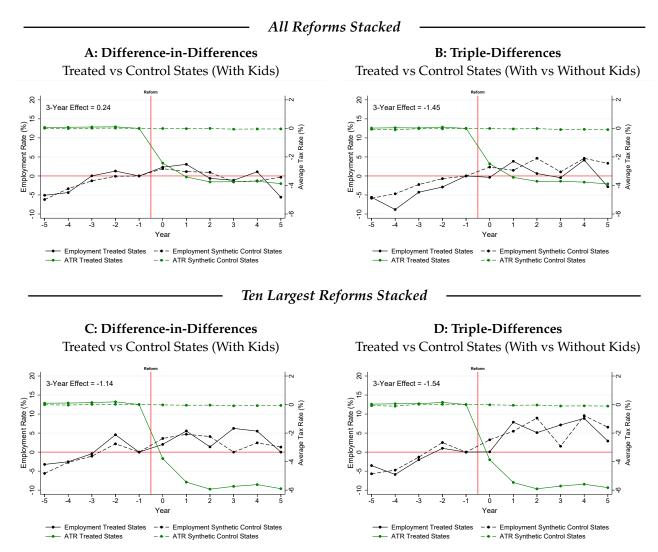


Ten Largest Reforms Stacked



Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is annual employment. The low-earnings sample includes single women in the bottom half of the within-year and within-state distribution of predicted earnings, estimated using equation (1).





Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is annual employment. The low-earnings sample includes single women in the bottom quartile of the within-year and within-state distribution of predicted earnings, estimated using equation (1).

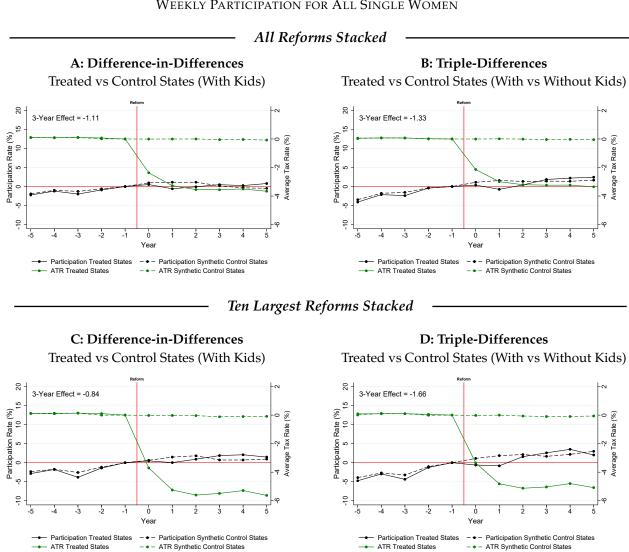
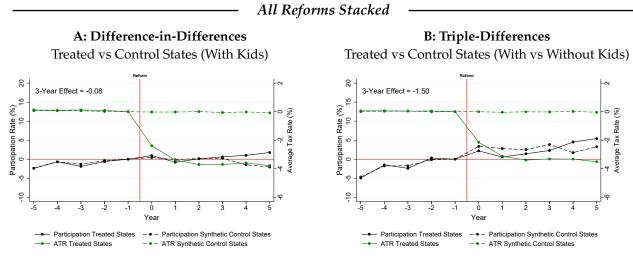


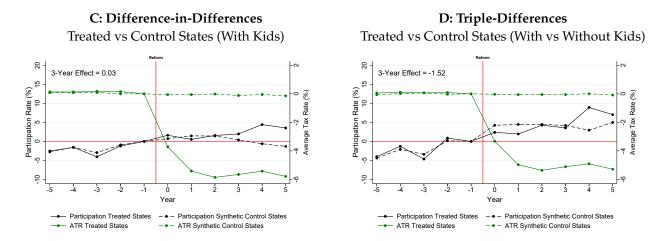
FIGURE A.90: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS WEEKLY PARTICIPATION FOR ALL SINGLE WOMEN

Notes: This figure is constructed exactly as Figure 11 in the main text, except for the extensive margin measure. In this figure, the sample is all single women and the extensive margin measure is weekly participation.

FIGURE A.91: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS WEEKLY PARTICIPATION FOR LOW-EDUCATED SINGLE WOMEN



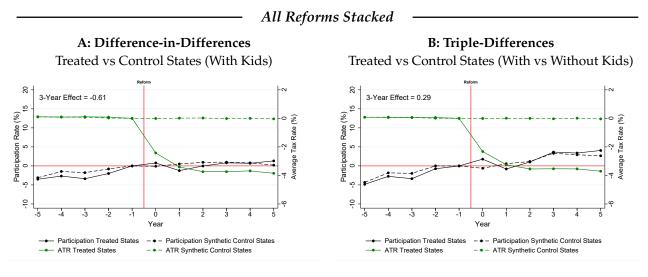




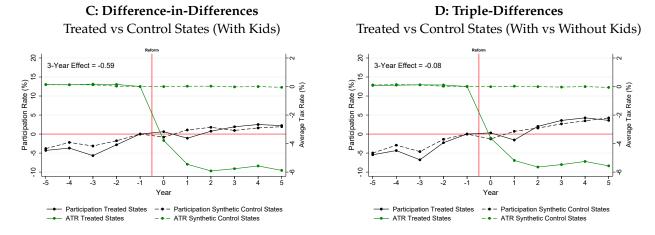
Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is low-educated single women (those with a high school degree or less) and the extensive margin measure is weekly participation.

FIGURE A.92: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS

WEEKLY PARTICIPATION FOR SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS

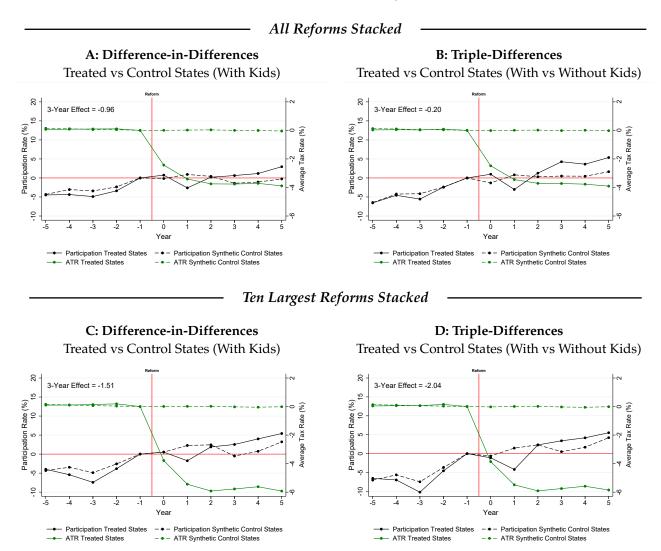


Ten Largest Reforms Stacked



Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is weekly participation. The low-earnings sample includes single women in the bottom half of the within-year and within-state distribution of predicted earnings, estimated using equation (1).

FIGURE A.93: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS WEEKLY PARTICIPATION FOR SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS



Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is weekly participation. The low-earnings sample includes single women in the bottom quartile of the within-year and within-state distribution of predicted earnings, estimated using equation (1).

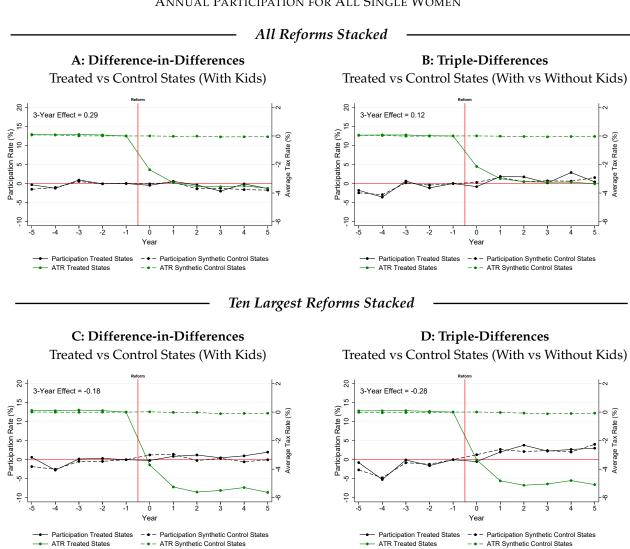
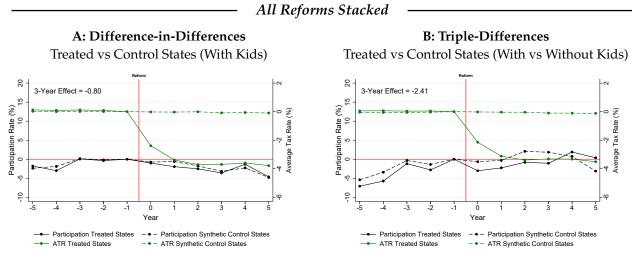


FIGURE A.94: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS ANNUAL PARTICIPATION FOR ALL SINGLE WOMEN

Notes: This figure is constructed exactly as Figure 11 in the main text, except for the extensive margin measure. In this figure, the sample is all single women and the extensive margin measure is annual participation.

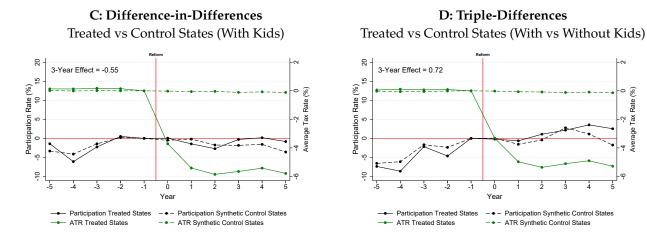
FIGURE A.95: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS ANNUAL PARTICIPATION FOR LOW-EDUCATED SINGLE WOMEN





%

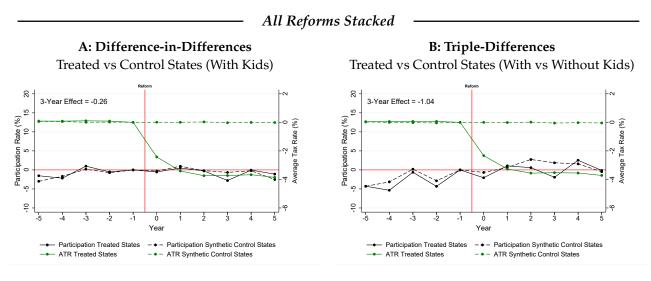
Average Tax Rate



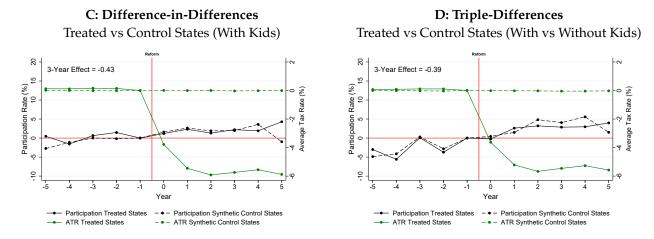
Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is low-educated single women (those with a high school degree or less) and the extensive margin measure is annual participation.

FIGURE A.96: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS

ANNUAL PARTICIPATION FOR SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS

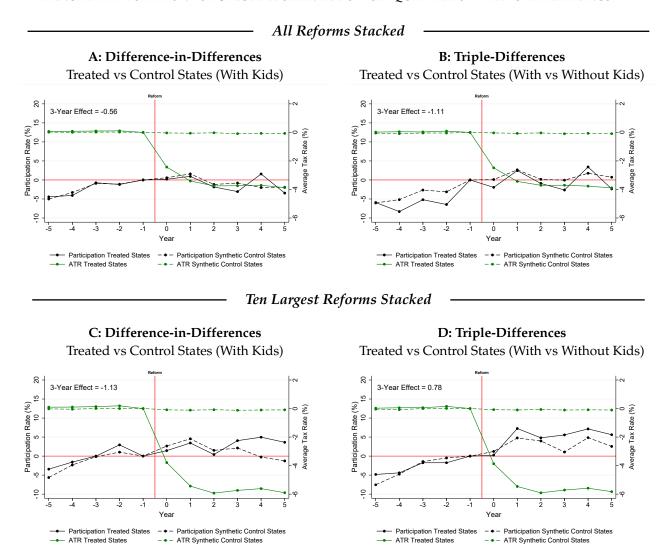


Ten Largest Reforms Stacked



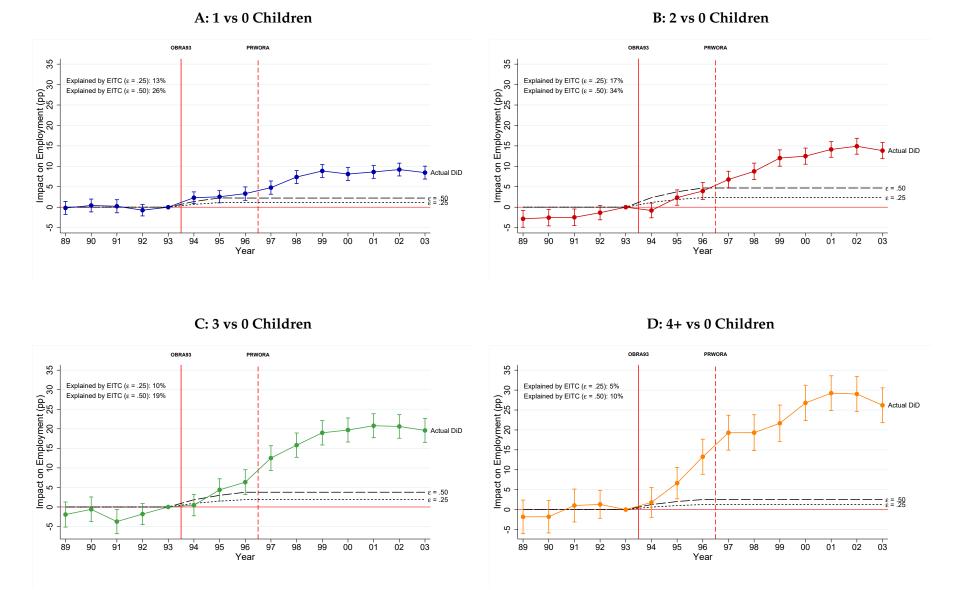
Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is annual participation. The low-earnings sample includes single women in the bottom half of the within-year and within-state distribution of predicted earnings, estimated using equation (1).

FIGURE A.97: SYNTHETIC CONTROL STUDY OF STATE EITC REFORMS ANNUAL PARTICIPATION FOR SINGLE WOMEN IN BOTTOM QUARTILE OF PREDICTED EARNINGS



Notes: This figure is constructed exactly as Figure 11 in the main text, except for the estimation sample and the extensive margin measure. In this figure, the sample is single women with low predicted earnings and the extensive margin measure is annual participation. The low-earnings sample includes single women in the bottom quartile of the within-year and within-state distribution of predicted earnings, estimated using equation (1).

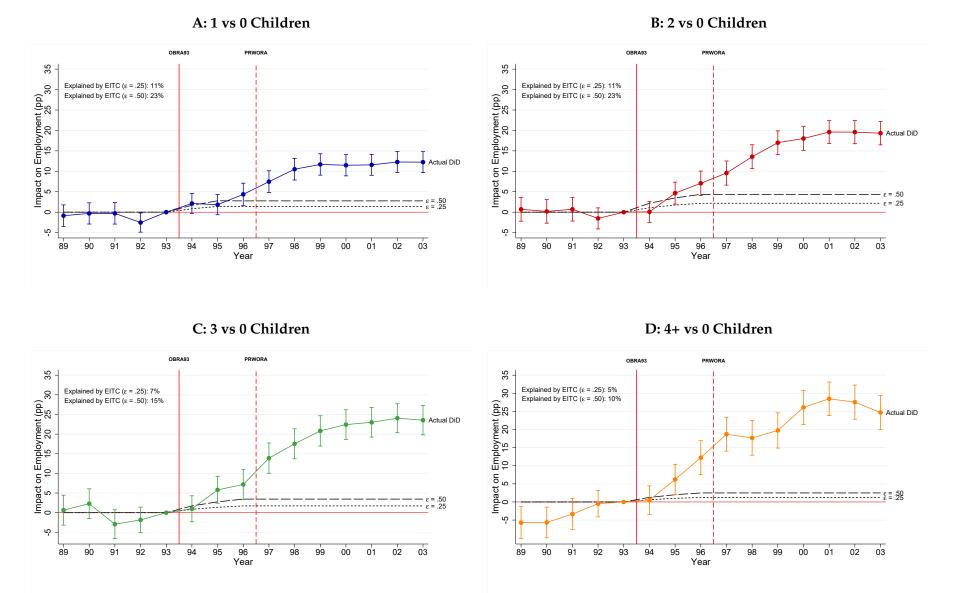
FIGURE A.98: ESTIMATED VS SIMULATED EFFECTS OF THE 1993 EITC REFORM All Single Women, Weekly Employment



Notes: This figure shows actual and simulated DiD event studies of the 1993 reform, by number of EITC-eligible children. The actual DiD series plot estimates $\hat{\gamma}_t^n$ based on an extension of specification (2) with separate dummies for each number of children *n* (and without any controls). The simulated DiD series plot ΔP_t calculated from equation (4), assuming an elasticity of 0.25 (short dashes) or 0.5 (long dashes). Each panel also shows the fraction of the employment increase over the post-reform period that can be explained by the 1993 tax reform (including but not limited to the EITC expansion) under the two elasticity scenarios. The extensive margin outcome is weekly employment, and the sample includes all single women aged 20-50 using the March and monthly CPS files combined. The 95% confidence intervals are based on robust standard errors clustered at the individual level.

FIGURE A.99: ESTIMATED VS SIMULATED EFFECTS OF THE 1993 EITC REFORM

SINGLE WOMEN IN BOTTOM HALF OF PREDICTED EARNINGS, WEEKLY EMPLOYMENT



Notes: This figure shows actual and simulated DiD event studies of the 1993 reform, by number of EITC-eligible children. The actual DiD series plot estimates $\hat{\gamma}_t^n$ based on an extension of specification (2) with separate dummies for each number of children *n* (and without any controls). The simulated DiD series plot ΔP_t calculated from equation (4), assuming an elasticity of 0.25 (short dashes) or 0.5 (long dashes). Each panel also shows the fraction of the employment increase over the post-reform period that can be explained by the 1993 tax reform (including but not limited to the EITC expansion) under the two elasticity scenarios. The extensive margin outcome is weekly employment, and the sample includes single women aged 20-50 with low predicted earnings using the March and monthly CPS files combined. The low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1). The 95% confidence intervals are based on robust standard errors clustered at the individual level.

171

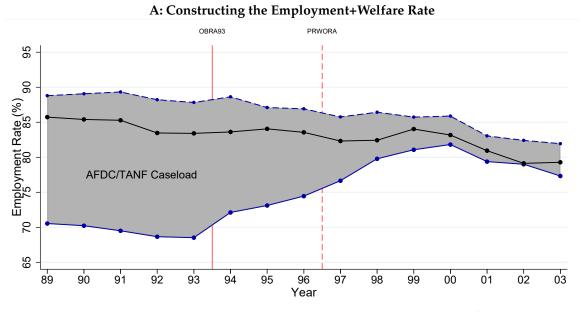
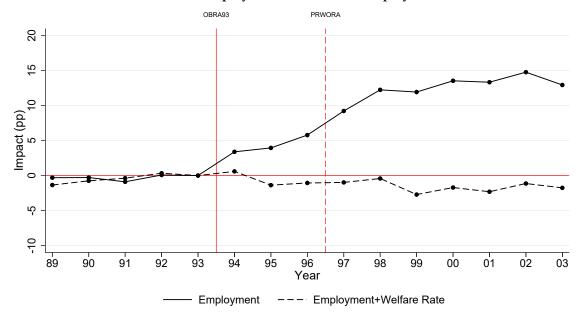


FIGURE A.100: EVENT STUDY IN THE EMPLOYMENT+WELFARE RATE

Employment with Kids — Employment without Kids - - Employment+Welfare Rate with Kids

B: Event Studies in the Employment Rate and the Employment+Welfare Rate



Notes: This figure shows the construction of the "employment+welfare rate" (Panel A) and event studies of the 1993 reform in the employment rate and the employment+welfare rate (Panel B). The employment+welfare rate is defined as the fraction of individuals who are employed and/or receiving AFDC/TANF benefits. Single women without children are ineligible for AFDC/TANF, and hence there is no distinction between the employment and the employment+welfare rate for this group. The outcome variables are measured at the annual level. The sample includes single women aged 20-50 using the March CPS files.

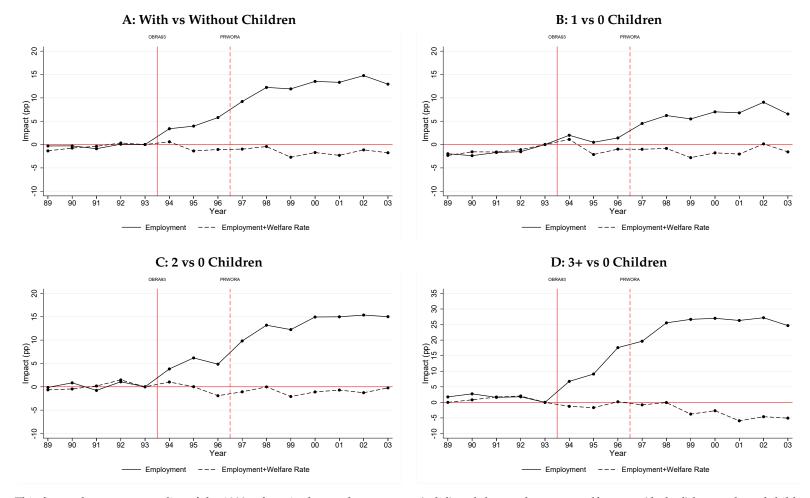


FIGURE A.101: EVENT STUDIES IN THE EMPLOYMENT+WELFARE RATE BY NUMBER OF CHILDREN

Notes: This figure shows event studies of the 1993 reform in the employment rate (solid) and the employment+welfare rate (dashed) by number of children. The employment+welfare rate is defined as the fraction of individuals who are employed and/or receiving AFDC/TANF benefits. The outcome variables are measured at the annual level. The sample includes single women aged 20-50 using the March CPS files.

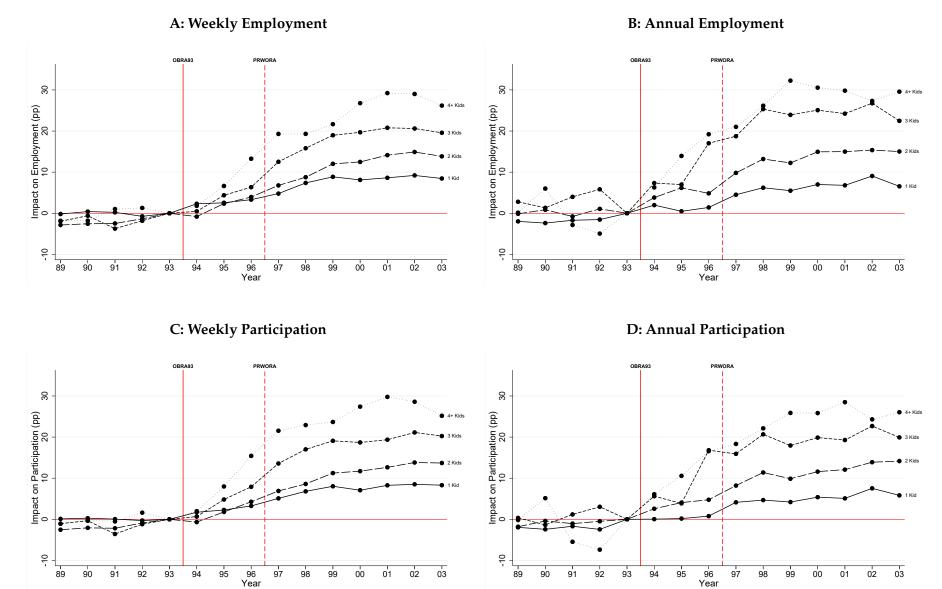
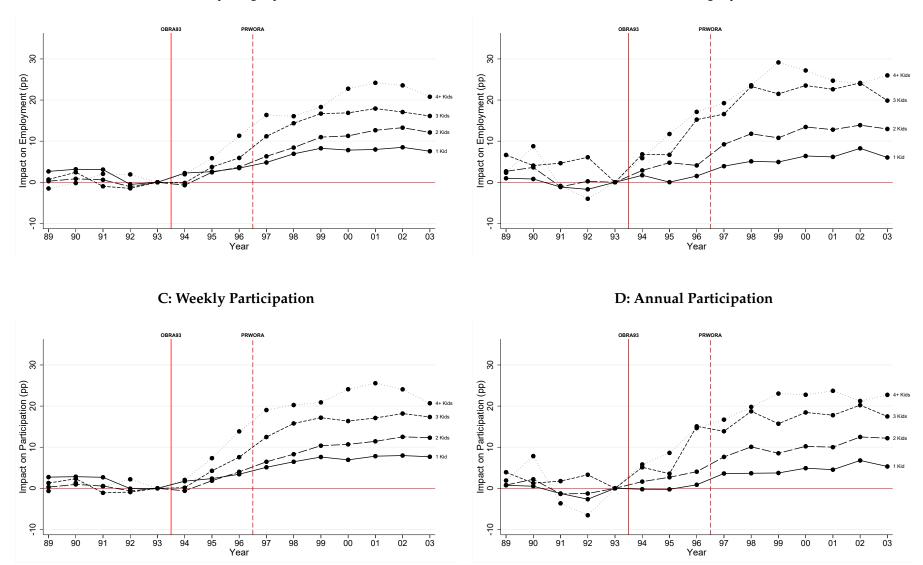


FIGURE A.102: FANNING OUT BY NUMBER OF CHILDREN All Single Women, Raw Data

Notes: This figure is constructed exactly as Panel A of Figure 14 in the main text, but shows results for all four extensive margin measures.

FIGURE A.103: FANNING OUT BY NUMBER OF CHILDREN All Single Women, With Demographic Controls

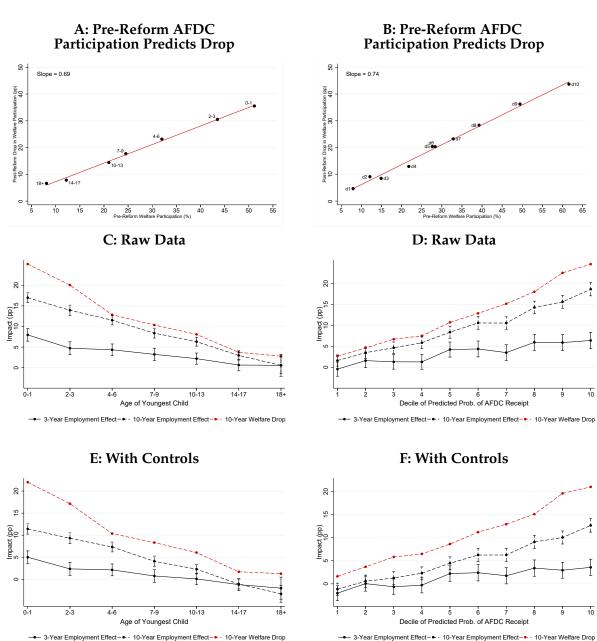


A: Weekly Employment

B: Annual Employment

Notes: This figure is constructed exactly as Panel B of Figure 14 in the main text, but shows results for all four extensive margin measures.

FIGURE A.104: EFFECTS OF THE 1993 EITC REFORM BY WELFARE TREATMENT INTENSITY All Single Women, Weekly Employment



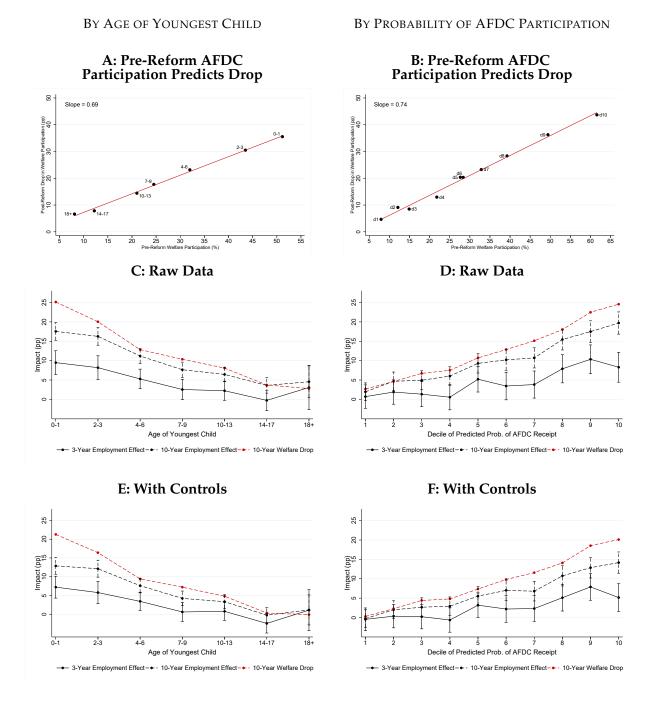
BY AGE OF YOUNGEST CHILD

BY PROBABILITY OF AFDC PARTICIPATION

176

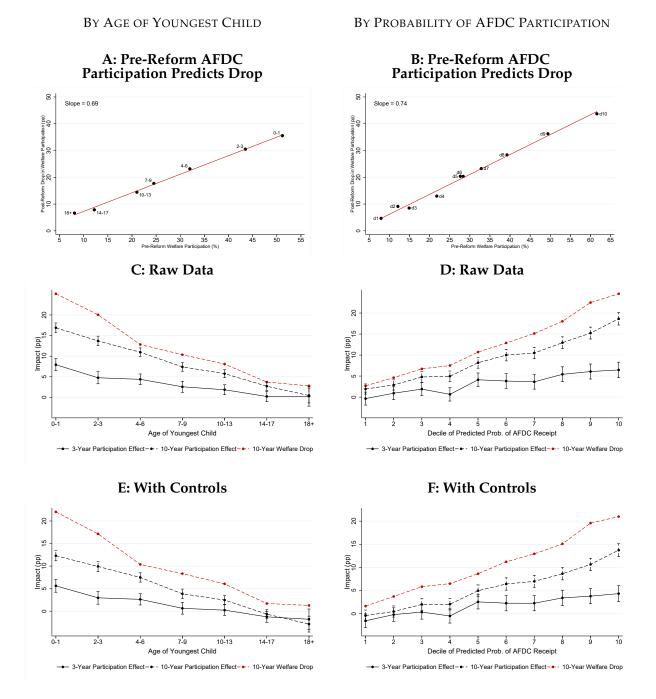
Notes: This figure is exactly the same as Figure 15 in the main text.

FIGURE A.105: EFFECTS OF THE 1993 EITC REFORM BY WELFARE TREATMENT INTENSITY All Single Women, Annual Employment



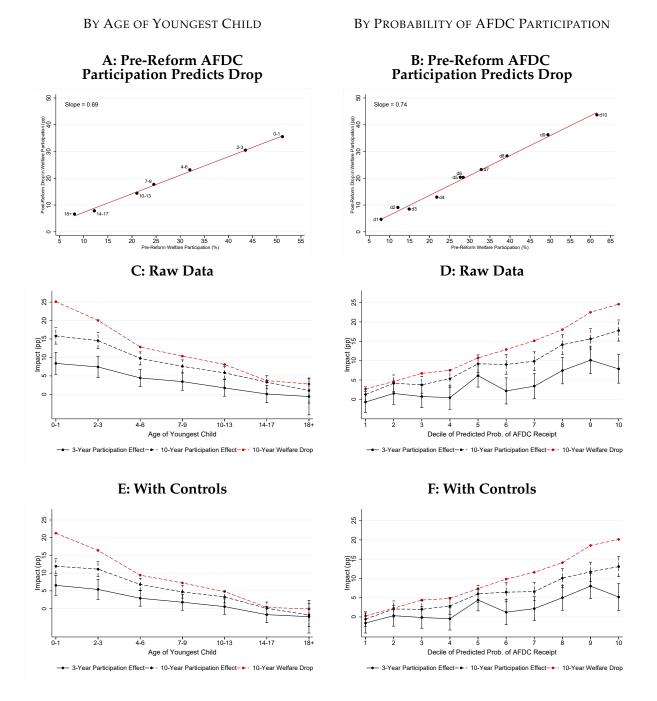
Notes: This figure is constructed exactly as Figure 15 in the main text, except that it uses annual employment as the extensive margin measure.

FIGURE A.106: EFFECTS OF THE 1993 EITC REFORM BY WELFARE TREATMENT INTENSITY All Single Women, Weekly Participation



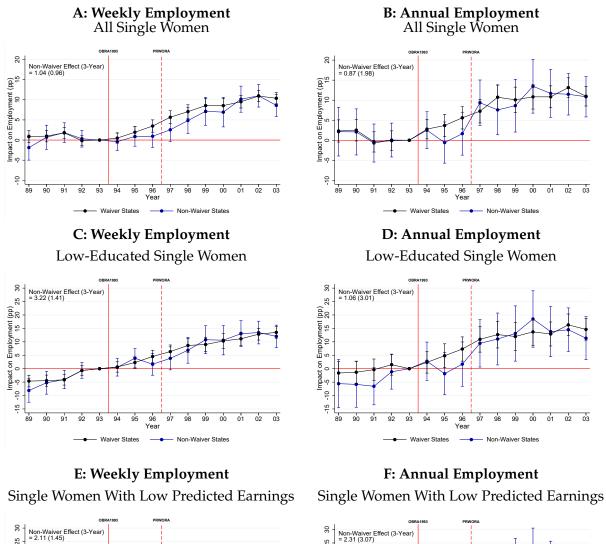
Notes: This figure is constructed exactly as Figure 15 in the main text, except that it uses weekly participation as the extensive margin measure.

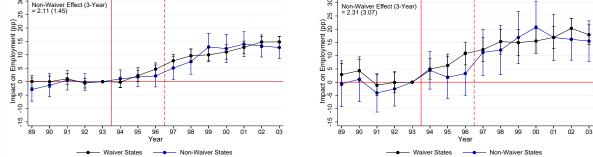
FIGURE A.107: EFFECTS OF THE 1993 EITC REFORM BY WELFARE TREATMENT INTENSITY All Single Women, Annual Participation



Notes: This figure is constructed exactly as Figure 15 in the main text, except that it uses annual participation as the extensive margin measure.

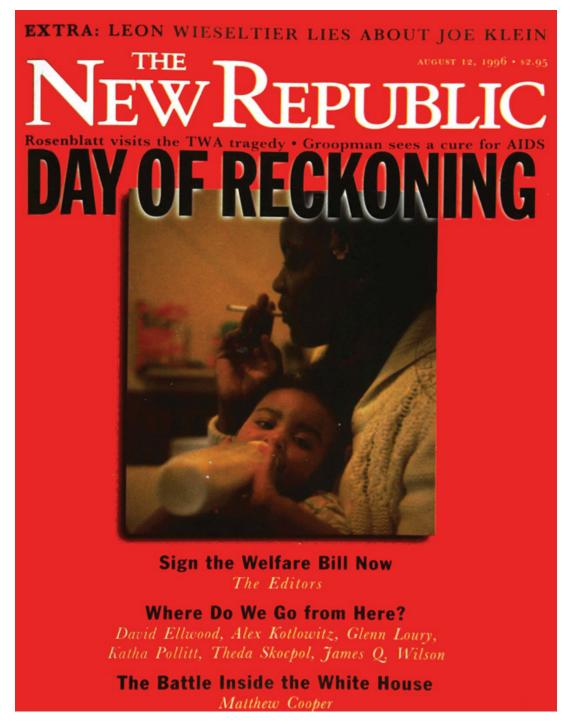






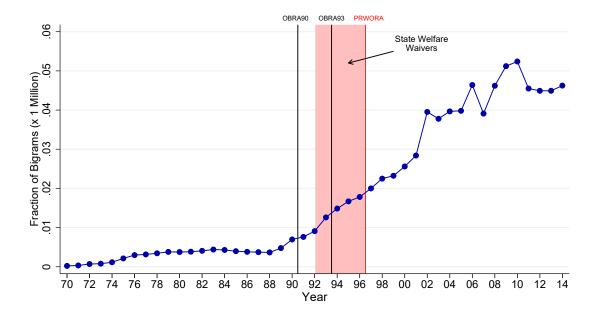
Notes: This figure is constructed exactly as Figure 16 in the main text, but shows results for two extensive margin measures (weekly and annual employment) and three estimation samples (all single women, low-educated single women, and single women with low predicted earnings). The low-education sample includes those with a high school degree or less, while the low-earnings sample includes single women in the bottom half of the within-year distribution of predicted earnings, estimated using equation (1).

FIGURE A.109: A PICTURE IS WORTH A THOUSAND WORDS



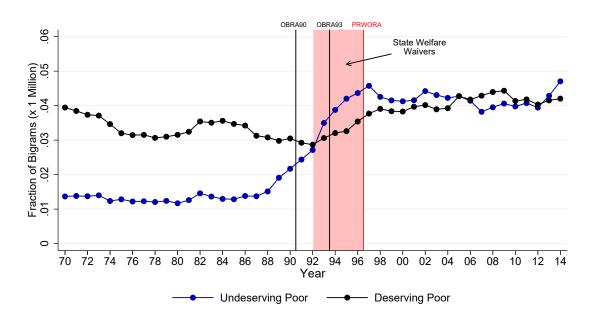
Notes: This figure shows the cover of the magazine *The New Republic* on August 12, 1996, just before PRWORA was signed into law by President Bill Clinton. It provides an example of the stereotypes and stigma surrounding welfare receipt in the 1990s.

FIGURE A.110: CHANGING SOCIAL NORMS



A: Use of "Welfare Queen" in US Publications

B: Use of "Undeserving Poor" and "Deserving Poor" in US Publications



Notes: This figure shows the usage of phrases "Welfare Queen," "Undeserving Poor," and "Deserving Poor" in books published in the US using Google Ngrams. The graphs show the five year moving average of the frequency of each phrase, which is measured as the number of occurrences of the phrase in each year, divided by the total number of bigrams in that year from the database of books digitized by Google.

B Data Description

B.1 Current Population Survey (CPS)

The CPS is made up of two main components: the Basic Monthly Survey and topical Supplements. In most cases, supplement samples are limited to individuals who participate in the Basic Monthly Survey. The Annual Social and Economic Supplement (ASEC) — the "March files" is an exception. It includes an oversample of respondents from other months who are not scheduled to receive the March Basic Monthly CPS. The ASEC is the most commonly used supplement of the CPS due to its long history, large sample size, and detailed information on annual income and social assistance. I use the Basic Monthly and the ASEC CPS files, extracted from IPUMS at https://cps.ipums.org/cps/.

The Basic Monthly CPS uses a sample rotation scheme whereby households are included in the sample for four consecutive months, excluded for eight, and then return for another four months before leaving the sample permanently. Due to the 4-8-4 sampling pattern, individuals in the CPS show up a total of eight times over a 16 month period. Despite this panel element of the CPS, most researchers use the survey as a repeated cross-section. While it is impossible to link respondents between the Basic Monthly samples and the ASEC oversample, it is possible to link respondents across the monthly samples alone. IPUMS has greatly simplified this process by creating the variable CPSIDP. CPSIDP is a combination of the year a household enters the sample, the month a household enters the sample, a within-month household ID, and a within-month person ID. It allows users to uniquely identify and track respondents across all Basic Monthly samples. CPSIDP is not available for respondents in the ASEC oversample, however, and it is therefore not possible to link respondents from the ASEC oversample to their observations in the Basic Monthly Survey.⁶⁶ More detail about how the unique ID is constructed given these constraints is described in section B.1.4.

I use the Basic Monthly files from 1989-2019 and the ASEC files from 1968-2019.⁶⁷ Although the monthly files are available from 1976, they do not allow for accurate identification of the presence and number of children in a household prior to 1989. To identify children in the CPS, I rely on the IPUMS variable RELATE. For each observation in a household, this variable identifies the relationship to the household head. Prior to 1989, the only RELATE categories available in the

⁶⁶See Flood and Pacas (2017) for a more comprehensive explanation for why the ASEC oversample respondents cannot be linked to their observations in the other months.

⁶⁷I exclude the March Basic file, because all respondents in the March Basic sample are included in ASEC.

monthly files are "householder," "spouse," "other relative," and "other non-relative." The ASEC files, on the other hand, have more consistent categories for RELATE over time and, importantly, these categories include "child." The absence of the "child" category in the monthly files results in a substantial undercounting of children relative to the ASEC files. Hence, I use the Basic Monthly files (combined with ASEC files) from 1989 onwards, and the ASEC files on their own before this time.

B.1.1 Extensive Margin Measures

I consider all four extensive margin measures available in the CPS: weekly employment, weekly participation, annual employment, and annual participation. Weekly measures are based on respondents' activities during the last week and are available in all Basic Monthly and ASEC files. Annual measures are based on respondent's activities during the last year and are only available in the ASEC.

Weekly Measures: Individuals' weekly employment and participation statuses are determined on the basis of answers to a series of questions relating to their activities during the preceding week. Upon answering these questions, respondents are grouped into eleven categories: "armed forces," "at work," "has a job, not at work last week," "unemployed, experienced worker," "unemployed, unexperienced worker," "housework," "unable to work," "school," "other," "unpaid, less than 15 hours," and "retired." Respondents classified as "at work" include those who either did any work for pay or profit or worked for at least fifteen hours without pay in a family business or farm. Respondents classified as "has a job, not at work last week" include those who did not work during the previous week but who acknowledged having a job or business from which they were temporarily absent (e.g. due to illness, vacation, or labor dispute). Individuals who do not fall into the above two categories but who reported either being temporarily laid off or actively searching for work are classified as unemployed.⁶⁸ Respondents who do not fall into any of the above categories are classified as not in labor force and distributed among the remaining six categories: "housework," "unable to work," "school," "other," "unpaid, less than 15 hours," and "retired."

Annual Measures: These are determined on the basis of questions in the ASEC pertaining to respondents' activities last year. The annual measures of employment and participation are based on different questions than the weekly measures. Annual employment is determined based on

⁶⁸Respondents were considered to be actively searching for work if they were either looking for work as their major activity during the previous week (for 1962 through 1993) or answered yes to a question about whether they had been looking for work in the past four weeks (for 1994 onwards).

respondents' earnings last year. Respondents with positive earnings are classified as employed last year. Annual participation is based on the number of weeks a respondent was either working or searching last year. Respondents who either worked or were looking for work for one or more weeks last year are classified as having participated last year.

B.1.2 Historical Changes to the CPS

In January 1994, the questions regarding labor force status (which underlie the weekly measures of employment and participation) underwent certain changes. A primary motivation for this redesign was to better classify individuals engaged in informal or intermittent activities. The redesign included a number of changes, all of which are explained in detail in Cohany, Polivka, and Rothgeb (1994). I focus here on the changes most pertinent to my analysis.

Prior to 1994, respondents were asked an "ice-breaker question" about their main activity during the preceding week. The question took the form "what were you doing most of last week? were you keeping house/working/in school or something else?" where the choice of prompt depended on the respondent's age and sex. The Bureau of Labor Statistics (BLS) concluded that this question led to an underreporting of women in part-time work. Additionally, respondents who indicated that they did not have a job were asked a follow-up question of the form "why were you absent from work last week?" Due to its open-ended nature, this question may have led to underreporting of respondents who were temporarily laid off. Beginning in 1994, these questions were redesigned to have more specific wording and fewer open-ended responses. For example, the initial "ice-breaker question" was replaced with a question asking if the respondent did any work for pay or profit last week. Similarly, respondents who indicated they did not have a job were asked whether they were laid off and if the layoff was temporary.

To asses the impact of this redesign on estimates of labor force participation, the BLS ran a parallel survey from July 1992 through December 1993 that interviewed households using the new survey questions. Cohany, Polivka, and Rothgeb (1994) examine the differences between official CPS and parallel survey estimates in a variety of metrics. They find that for women aged 20 and above, the weekly employment rate was 55.1% in the official CPS and 55.8% in the parallel survey, a difference of only 0.7 percentage points. The weekly participation rate was 58.5% in the official CPS and 59.6% in the parallel survey, a difference of 1.1 percentage points. When including controls for state of residence, race, and Hispanic origin, these differences drop to 0 and 0.1 percentage points, respectively. These differences are too small to have any substantial effect on the analysis. In any

case, the difference-in-differences design reduces this issue even further, or eliminates it entirely, by including year fixed effects.

B.1.3 Nonresponse in the CPS

The CPS is subject to two types of nonresponse: noninterview households and item nonresponse. Noninterview occurs when a household refuses to participate in the survey altogether and is especially common in March, corresponding to the delivery of the ASEC. In the Basic Monthly CPS, noninterview is accounted for by distributing the weights of noninterview households among interviewed households. In the ASEC, noninterview is accounted for by imputing missing values.

The second source of data loss, item nonresponse, occurs when respondents refuse to answer specific questions within the survey. To compensate for item nonresponse (and for noninterview in the ASEC), the BLS imputes missing values using one of three methods. First, if possible, missing values are inferred from other characteristics of a respondent or other respondents within the same household. For example, if a respondent has a missing value for race, it is assigned based on the race of other household members. These edits, known as relational edits, are most commonly used for demographic variables. Next, if relational edits are not possible, longitudinal edits are made. Longitudinal edits exploit the panel nature of CPS data and use respondent's entries from previous months to fill in missing values. Labor force items are typically imputed using longitudinal edits. Finally, if neither of the above are possible, the CPS uses a "hot-deck" imputation method. The "hot-deck" method assigns a missing value from a record with similar characteristics, called the hot deck. Hot decks are made up of demographic characteristics such as age, race, sex, occupation and educational attainment. The specific characteristics that make up a hot deck vary depending on which variable is being imputed.

How common is nonresponse in the CPS? Historically, nonresponse in the CPS was very modest, but it has grown significantly over time (see e.g., Meyer, Mok, and Sullivan 2015; Jones and Ziliak 2022). Household non-interview rates have risen from 7-9 percent in 2004 to 13-15 percent in 2017.⁶⁹

As for item nonresponse, two points are worth mentioning. First, item nonresponse is much smaller for demographic and labor force variables than it is for earnings. In 2018, only 0.45% of the respondents in the estimation sample have imputed labor force status and 3.1% have imputed

⁶⁹These statistics have been retrieved from https://www.census.gov/prod/2006pubs/tp-66.pdf and https://www.census.gov/programs-surveys/cps/technical-documentation/methodology/non-response-rates.html.

demographics (marital status, age, or race), compared to 17.9% with imputed earnings. Second, the degree of item nonresponse in earnings has increased over time. While 17.9% of respondents have imputed earnings in 2018, this number was only 10.9% in 1970.⁷⁰ The significant degree of nonresponse and imputation in the earnings variable is another argument for using the weekly measures of extensive margin labor supply, as I do here.

B.1.4 Sample and Variables

I restrict the sample to single women (never married, separated, divorced, or widowed) aged 20- $50.^{71}$ I drop observations with a zero, negative, or missing weight (*wgt*), missing state FIPS code (*stfips*), or missing educational attainment (*educ*). The difference-in-differences analyses are based on comparing single women with EITC eligible children (treatment group) to single women without recorded children (control group). The control group includes both those who never had any children and those whose children do not live at home. A small fraction of single women with EITC *ineligible* children living at home are dropped from the sample.⁷² These restrictions leave me with a sample of 4,809,195 individual-month observations across survey years 1968-2019.

Unless otherwise specified, variables in the ASEC and monthly files are defined in the same way. Variables based on income and welfare participation are only available in the ASEC.

- Unique household ID (*hid*): This variable is my best attempt at a unique identifier for each household in the CPS. In the monthly files, households can be uniquely identified and tracked across subsequent months using IPUMS variable CPSID. In the ASEC files, the variable CP-SID is unavailable so I instead identify households using a combination of IPUMS variables YEAR and SERIAL. As a result, respondents in the ASEC cannot be linked to those in the monthly files and the variable *hhid* only uniquely identifies households in the monthly and ASEC files separately.
- Unique person ID (*id*): This variable is a unique combination of *hhid* and IPUMS variable PERNUM, which uniquely identifies individuals within a household.

⁷⁰The growth in item nonresponse rates has also been quite large for other income variables, including income from social assistance programs. See Meyer, Mok, and Sullivan (2015) and Meyer and Mittag (2019) for an investigation of item nonresponse bias in questions pertaining to social assistance receipt.

⁷¹Except for one analysis in the online appendix in which I consider a sample of all women.

⁷²For example, this includes women with children who recently turned 19 and are not full-time students. The reason for dropping these observations (as opposed to assigning them to the control group) is that most of them would have been EITC eligible in the recent past and are therefore borderline cases between the treatment and control groups. In any case, assigning them to the control group does not make much of a difference to any of the results.

- Number of eligible children (*nechild*): This variable identifies the number of EITC eligible children a respondent has. An EITC eligible child is defined as a household member who is either under 19 or who is under 24 and a full time student (EMPSTAT = 33). Using a combination of household ID (*hhid*) and IPUMS variables MOMLOC and POPLOC, I link respondents to their biological and adoptive children. I then look to the age and education associated with each child's observation to establish whether the child is EITC eligible. For more detail on how to link respondents to their children, see https://cps.ipums.org/cps-action/variables/MOMLOC#description_section.
- Age of youngest child (*ageyec*): The minimum age of all EITC eligible children. Takes on a value of 99 if respondent has no children.
- Single (*single*): Takes on a value of one if the respondent is separated (MARST = 3), divorced (MARST = 4), widowed (MARST = 5), or never married (MARST = 6), a value of zero if the respondent is married with spouse present (MARST = 1) or married with spouse absent (MARST = 2), and is missing otherwise.
- Age (*age*): this variable is taken from the IPUMS *age* variable and is top-coded at 90.
- Gender (*female*): takes on a value of one if IPUMS variable SEX = 2 and zero otherwise.
- Education Level (*edlevel*): takes on a value of one if the respondent has less than a HS education (IPUMS variable EDUC = 2-72), a value of two if the respondent has a HS diploma or equivalent (EDUC = 73), and a value of three if the respondent has more than a HS education (EDUC = 80-125).
- Low-educated (*lowed*): takes on a value of one if respondent has less than a HS education (*edlevel* = 1 − 2) and zero otherwise.
- Alternate low-educated (*lowedB*): takes on a value of one if respondent has less than a HS education (*edlevel* = 1) and zero if the respondent has HS education and above (*edlevel* = 2-3).
- Race (*raced*): takes on a value of one if the respondent is white (IPUMS variable RACE = 100), a value of two if the respondent is black (RACE = 200), a value of 3 if the respondent is Asian/Pacific islander (RACE = 650-652), and a value of four otherwise (RACE = 300 and RACE = 700-830).

- AFDC receipt (*afdc_annual*): takes on a value of one if respondent receives AFDC/TANF (SRCWELFR = 1) or both AFDC/TANF and another type of welfare (SRCWELFR = 3), takes on a value of zero if respondent doesn't receive welfare (SRCWELFR = 0) or receives only another type of welfare (SRCWELFR = 2), and is missing otherwise.
- Weekly employment (*emp*): takes on a value of one if respondent is in the armed forces (EMPSTAT = 1), working (EMPSTAT = 10), or has a job but is not at work (EMPSTAT = 12), a value of zero if respondent is unemployed (EMPSTAT = 20-22), or not in the labor force (EMPSTAT = 30-36), and is missing otherwise.
- Weekly participation (*lfp*): takes on a value of one if the respondent is in the armed forces (EMPSTAT = 1), working (EMPSTAT = 10), has a job but is not at work (EMPSTAT = 12), or is unemployed (EMPSTAT = 20-22), a value of zero if the respondent is not in the labor force (EMPSTAT = 30-36), and is missing otherwise.
- Annual employment (*emp_annual*): takes on a value of one if person had positive earnings last year (IPUMS variable INCWAGE > 0), zero if they had zero earnings last year.⁷³
- Annual participation (*lfp_annual*): takes on a value of one if the respondent worked (WKSWORK1) or looked for work (NWLOOKWK) for at least one week last year and takes on a value of zero if the respondent didn't look for work at all last year (WKSWORK1 = 0 and NWLOOKWK = 0), and is missing otherwise.
- Income (*wsal*): the *wsal* variable comes from the IPUMS variable INCWAGE. Values of 9999999 and 9999998 are recoded to be missing.
- Person weight (*wgt*): in the ASEC this variable is equal to the IPUMS variable ASECWT; in the monthly files this variable is equal to the IPUMS variable WTFINL.
- State unemployment rate (*st_unemployed*): the state unemployment rate is calculated by dividing the number of unemployed respondents (EMPSTAT = 20-22) by the number of respondents in the armed forces (EMPSTAT = 1), working (EMPSTAT = 10), or with a job but not at work (EMPSTAT = 12) in a given state.

⁷³When using the annual employment variables (lfp_annual and emp_annual) as outcomes, I substitute *year* for year - 1 to reflect the fact that the employment measure refers to the previous year.

B.2 Supplementary Data

Data on state welfare waivers comes from the Department of Health and Human Services (HHS).⁷⁴ I follow HHS and consider major statewide waivers in the following six categories: termination time limits, work requirement time limits, JOBS exemptions, JOBS sanctions, family caps, and earnings disregards.

Data on federal EITC parameters come from the Tax Policy Center.⁷⁵ Data on state EITC parameters come from the Tax Policy Center,⁷⁶ the National Bureau of Economic Research (NBER),⁷⁷ and various state-specific sources.

C Tax Simulations

I use NBER's tax simulation model (TAXSIM) to calculate the average tax rate (ATR) relevant for extensive margin labor supply decisions. The average tax rate includes tax liabilities from state income taxes, federal income taxes, and payroll taxes. It is defined as

$$ATR(Y) = \frac{(T_f(Y) - T_f(0)) + (T_s(Y) - T_s(0)) + T_p(Y)}{Y},$$
(7)

where *Y* is earnings conditional on working, $T_f(Y)$ and $T_f(0)$ ($T_s(Y)$ and $T_s(0)$) are the federal (state) income tax liabilities when working and not working, respectively, and $T_p(Y)$ is the payroll tax liability (Federal Insurance Contributions Act tax) when working.

TAXSIM requires information on income variables, dependents, and demographics.⁷⁸ The following list describes the mapping between TAXSIM variables (shown in parentheses) and CPS variables:

- Marital status (*mstat*) is set as "single or head of household" (corresponding to a value of one) for all observations.
- Age (*page*) is equal to the variable *age* described in appendix section B.1.4.

⁷⁴Health and Human Services, Assistant Secretary for Planning and Evaluation, State Implementation of Major Changes to Welfare Policies, 1992-1998. Retrieved from https://aspe.hhs.gov/report/state-implementation-major-changes-welfare-policies-1992-1998.

⁷⁵Tax Policy Center. "Earned Income Tax Credit Parameters, 1975-2018." Retrieved from https://www.taxpolicycenter.org/statistics/eitc-parameters.

⁷⁶Tax Policy Center. "State EITC as Percentage of the Federal EITC" Retrieved from https://www.taxpolicycenter.org/statistics/state-eitc-percentage-federal-eitc.

⁷⁷NBER. "State EITC provisions 1977-2016." Retrieved from https://users.nber.org/~taxsim/state-eitc.html.

⁷⁸The full list of TAXSIM inputs is listed online at https://users.nber.org/~taxsim/taxsim32/

- Number of dependents (*depx*) uses IPUMS variable NCHILD, which corresponds to the number of own children at home.
- Number of children under 13 with eligible child care expenses (*dep*13) is equal to number of children at home that are under 13. Uses variables NCHILD (described above) and *ageyec* (described in appendix section B.1.4).
- Number of children under 17 for the entire tax year (*dep*17) is equal to number of children at home that are under 17. Uses variables NCHILD (described above) and *ageyec* (described in appendix section B.1.4).
- Number of qualifying children for EITC (*dep*18) is equal to variable *nechild* (described in appendix section B.1.4).
- Earnings from wages and salary (*pwages*) is either zero (when not working) or *Y* (when working). As will be described below, I use two different approaches to obtain *Y* from the CPS data.
- Dividends (*dividends*) comes from IPUMS variable INCDIVID. INCDIVID indicates how much pre-tax income the respondent received from stocks and mutual funds.
- Interest received (*intrec*) comes from IPUMS variable INCINT. INCINT indicates how much pre-tax income the respondent received from interest on saving accounts, certificates of deposit, money market funds, bonds, treasury notes, IRAs, and/or other investments which paid interest.
- Other property income (*otherprop*) comes from IPUMS variable INCRENT. INCRENT indicates how much pre-tax income the respondent received from rent (after expenses), from charges to roomers or boarders, and from money paid by estates, trusts, and royalties.
- Gross Social Security benefits (*gssi*) comes from IPUMS variable INCSS. INCSS indicates how much pre-tax income the respondent received from Social Security payments.
- Unemployment insurance (*ui*) comes from IPUMS variable INCUNEMP. INCUNEMP indicates how much pre-tax income the respondent received from state or federal unemployment compensation, Supplemental Unemployment Benefits (SUB), or union unemployment or strike benefits.

- Age and wage of spouse (*sage*, *swages*) are set to zero as the sample only includes single women.
- Other income (*stcg*, *ltcg*, *mortgage*, *nonprop*, *pensions*, *rentpaid*, *proptax*, *otheritem*, *childcare*) are set to zero as they are not observed in the CPS.

For each individual, tax liability is simulated when working (pwages = Y) and when not working (pwages = 0). The setting of earnings when working, Y, is done differently for the elasticity approximations (see section C.1 below) and the labor supply simulations (see section D.1 below).

C.1 Calculation of ATRs for Figure 2 and Elasticities

The calculation of ATRs requires a measure of earnings conditional on working for both workers and non-workers. For each woman observed in year t with n children, I set earnings when working equal to the first kink of the federal EITC in year t for families with n children. Prior to 1975, there was no EITC. Hence, for the years 1968-1974, I set earnings equal to the first kink of the federal EITC in 1975, adjusted for inflation. Prior to 1994, there was no EITC for families without children. For childless women during the years 1968-93, I set earnings equal to the first kink of the federal EITC for childless families in 1994, adjusted for inflation.

The resulting ATRs for single women with different numbers of children between 1968-2018 are plotted in Figure 2. In this figure, because TAXSIM only includes *state* tax laws from 1977 onwards, the calculation of ATRs prior to 1977 is based on state ATRs in 1977 within cells of number of children (0, 1, 2, 3+) and cells of education (4 categories).

The extensive margin elasticity with respect to the average net-of-tax rate is defined as

$$\varepsilon \equiv \frac{\Delta P/P}{\Delta \left(1-\tau\right) / \left(1-\tau\right)}.$$
(8)

The numerator of the elasticity is the percentage effect of a given tax reform on the extensive margin labor supply of single mothers. I focus on average three-year effects to avoid overlap between reforms. Specifically, ΔP is estimated as the difference-in-differences between treatment and control groups using a modified version of equation (2) in which the three year dummies just after the reform are collapsed into a single post dummy. The baseline employment or participation rate of single mothers *P* is calculated using data from the pre-reform year. The numerator $\Delta P/P$ is obtained by dividing the two population averages.

The denominator of the elasticity is the percentage effect of the tax reform on the average netof-tax rate. That is, the calculation of extensive margin elasticities is based on $\tau = ATR$, using the tax rate calculations shown in Figure 2. To be consistent with the numerator, I focus on the average three-year change in the ATR, calculated as the difference-in-differences between treatment and control groups using the following specification

$$ATR_{imt} = \alpha \cdot Post_t + \beta \cdot Kids_i + \gamma \cdot Post_t \cdot Kids_i + \nu_{imt}, \tag{9}$$

where $Post_t$ is a dummy variable equal to 1 in the years after the reform. The regression is run on a four-year sample, the pre-reform year and three post-reform years, ensuring that the time horizon is consistent with the numerator. The reform-induced change in the average net-of-tax rate for the treatment relative to control groups, $\Delta (1 - ATR)$, is given by $-\gamma$. The baseline tax rate is the ATR of single mothers in the pre-reform year. The denominator $\Delta (1 - ATR) / (1 - ATR)$ is obtained by dividing the two population averages.

D Benefit Simulations

To simulate the impact of the 1993 tax reform on extensive margin labor supply (as presented in section 7.1), I calculate participation tax rates that account for both taxes paid and benefits lost when entering the labor market. To this end, I combine NBER's tax simulation model (TAXSIM) with a benefit calculator that includes Aid to Families with Dependent Children (AFDC) and Food Stamps (FS). The participation tax rate (PTR) is defined as

$$PTR(Y) = ATR(Y) + takeup \times \frac{B(0) - B(Y)}{Y},$$
(10)

where ATR(Y) is the average tax rate defined in equation (7), *takeup* is the average take-up rate of benefits by eligibles, while B(0) and B(Y) are the annual benefits from AFDC and Food Stamps when not working and working, respectively.

Annual benefits from AFDC and Food Stamps when not working (B(0)) are calculated as

$$B(0) = (AFDC(0) + FS(0)) \cdot 12$$

where AFDC(0) and FS(0) denote the maximum monthly benefits from AFDC and Food Stamps (in 1993), which vary by family size and by state.

Next, I calculate the annual benefits from AFDC and Food Stamps when working (B(Y)). For those who are working, benefits from AFDC (AFDC(1)) and Food Stamps (FS(1)) are clawed back as earned income increases. However, part of earned income is disregarded when determining benefit levels. Specifically, monthly benefits are calculated as follows

$$AFDC(1) = \min(PaymentStandard - NetIncome_{AFDC}, AFDC(0)) \cdot \mathbf{I}_{AFDC}$$
$$FS(1) = (FS(0) - 0.3 \cdot NetIncome_{FS}) \cdot \mathbf{I}_{FS},$$

where the indicators I_{AFDC} and I_{FS} are equal to one when the calculated AFDC and Food Stamp benefits are non-negative and equal to zero otherwise. These formulas account for the fact that *AFDC* has a phase-out rate of 1 (each additional dollar of net income implies \$1 less of *AFDC*), while *FS* has a phase-out rate of 0.3 (each additional dollar of net income implies \$0.3 less of food stamps). The variable *PaymentStandard* denotes a baseline AFDC benefit level (before claw-back) used to compute benefits for working individuals in 1993. The payment standard varies by family size and by state. It is typically equal to *AFDC*(0), but some states set *PaymentStandard* to a level above *AFDC*(0). The net income measures (*NetIncome_{AFDC}* and *NetIncome_{FS}*) are defined as follows

$$\begin{split} NetIncome_{AFDC} &= \max \left(GrossIncome - WorkExp - ChildCareExp \cdot n \\ -EarningsDisregard_{AFDC}, 0 \right) \\ NetIncome_{FS} &= \max \left(GrossIncome - ShelterCosts - ChildCareExp \cdot n \\ -StandardDeduction - EarningsDisregard_{FS}, 0 \right). \end{split}$$

The following list describes the components of the monthly benefit calculation:

• Gross Income (*GrossIncome*) is calculated by dividing annual wage and salary income (*Y*) by the number of months worked (*MonthsWorked*).⁷⁹ The variable *MonthsWorked* is calculated from the IPUMS variable WKSWORK1, which is the number of weeks worked. For non-workers, I impute a value for *MonthsWorked* based on the average value for workers in different cells of family size *n*.

⁷⁹In principle, Gross Income also includes unearned income (interest income, dividends, capital gains, etc.), but I set such income equal to zero in this sample of single women.

- Work Expenses Disregard (*WorkExp*) denotes work-related expenses such as transportation costs. *WorkExp* is set equal to the maximum disregard in 1993 (\$90).
- Child Care Expenses Disregard (*ChildCareExp*) denotes child care expenses incurred per child. *ChildCareExp* is set equal to the maximum disregard in 1993 (\$175 per child for AFDC and \$160 per child for Food Stamps).
- AFDC Earnings Disregard (*EarningsDisregard_{AFDC}*) denotes an earned income disregard, which provides AFDC recipients a financial incentive to seek and maintain employment. It is equal to \$30 + 0.33 · *Earnings* per month for the first 4 months worked and \$30 per month for all subsequent months.
- Food Stamps Earnings Disregard (*EarningsDisregard_{FS}*) denotes an earned income disregard, which provides Food Stamps recipients a financial incentive to seek and maintain employment. It is equal to 0.2 · *Earnings*.
- Standard Deduction (*StandardDeduction*) is a monthly deduction to account for basic costs, which varies by state.
- Shelter Deduction (*ShelterCosts*) is an allowance for excess housing costs. It is equal to housing expenses (such as rent and utilities) exceeding 50 percent of "net income," up to a maximum. Here, net income is computed as earnings (*Y*) net of the standard deduction (*StandardDeduction*) and the earnings disregard (*EarningsDisregard_{FS}*). It is assumed that households living in rented properties incur housing expenses equal to 30 percent of monthly earnings, whereas households living in owned properties are assumed to have no allowance for excess housing costs. Property ownership is determined using the IPUMS variable OWNERSHP, which indicates whether the household rented or owned its housing unit. The maximum shelter cost deduction varies by state.

The annual benefits from AFDC and Food Stamps are calculated as follows

$$B(Y) = AFDC(1) \cdot MonthsWorked + AFDC(0) \cdot (12 - MonthsWorked) +FS(1) \cdot MonthsWorked + FS(0) \cdot (12 - MonthsWorked),$$

where AFDC(1) and FS(1) are functions of earnings conditional on working Y through the net

and gross income measures that enter into their formulas as described above. Eligible individuals working some of the year receive benefits AFDC(1) and FS(1) in each month of working, and they receive the maximum benefits AFDC(0) and FS(0) in the remaining months.

The values of B(0) and B(Y) described above enter the participation tax rate formula (10). The average take-up rate *takeup* scales welfare benefits to account for incomplete benefit take-up among eligibles. Based on the evidence presented in Blank and Ruggles (1996), Blank (2002), and Currie (2006), the average take-up rate is set equal to 70%. The participation tax rate is top-coded at 0.99.

D.1 Calculations of PTRs used in Labor Supply Simulations

The calculation of participation tax rates requires information on earnings conditional on working, Y. Unlike the elasticity approximations described above (assuming single women enter the labor market at the first kink of the federal EITC), here I take an exact approach that accounts for the entire distribution of earnings. Specifically, I use observed earnings for workers and predicted earnings for non-workers, where the latter is estimated based on specification (1) in the main text:

$$Y_i = \alpha_a + \beta_n + \gamma_y + \delta_e + \zeta_r + \eta_s + \nu_i.$$

Earnings are regressed on dummies for the age of the woman a (20-24, 25-29, 30-34, 35-39, 40-44, 45-50), dummies for the number of EITC eligible children n (0, 1, 2, 3, 4, 5, 6+), dummies for the age of youngest child y (0-1, 2-3, 4-6, 7-9, 10-13, 14-17, 18+), dummies for education level e (below high school, high school degree, some college, college degree), dummies for race r (white, black, Asian, and other), and dummies for state of residence s. The regression is run on a pre-reform sample (1993) of single women with positive earnings. Based on the parameters from this regression, earnings are predicted for non-workers. Those with predicted earnings below zero are dropped (131 observations).

Having obtained a measure of earnings conditional on working for both workers and nonworkers, participation tax rates can be simulated using equation (10). The labor supply impact of the 1993 reform can then be simulated based on equation (4) in the main text:

$$\Delta P_t = \varepsilon \cdot \frac{\Delta \left(1 - \tau_t\right)}{1 - \tau_{93}} \cdot P_{93},$$

where ΔP_t is the impact on extensive margin labor supply in year *t* relative to the pre-reform year

1993, ε is the extensive margin elasticity, $\Delta (1 - \tau_t)$ is the reform-induced change in the net-of-tax rate on participation in year t relative to 1993, $\tau_{93} = PTR_{93}$ is the baseline participation tax rate in 1993, and P_{93} is the baseline employment rate in 1993. Because the goal is to simulate the impact of the federal tax reform act of 1993, the calculation of $\Delta (1 - \tau_t)$ accounts only for *federal income tax* changes (including but not restricted to the EITC expansion), holding the tax rates from federal payroll taxes, state income taxes, and welfare benefits constant at their 1993 levels. The federal income taxes in each year t are based on 1993 earnings adjusted for inflation, thereby isolating the mechanical effect of the tax reform. Calculating $\Delta (1 - \tau_t)$ as the difference-in-differences between treatment and control groups in each year t relative to the pre-reform year 1993, the implied time path of ΔP_t can be compared to the observed difference-in-differences impacts from the event studies.

E 1975 EITC Introduction: Triple-Differences Specification

Appendix Figure A.17 presents results from a triple-differences specification comparing treated and untreated women (in terms of their marital status and spousal earnings) with and without children over time. The estimates in the figure are based on the following specification:

$$P_{imt} = \sum_{j} \alpha_{j} \cdot Year_{j=t} + \beta \cdot Kids_{i} + \gamma \cdot Treat_{i} + \sum_{j \neq -1} \delta_{j} \cdot Year_{j=t} \cdot Kids_{i} + \sum_{j \neq -1} \zeta_{j} \cdot Year_{j=t} \cdot Treat_{i} + \eta \cdot Kids_{i} \cdot Treat_{i} + \sum_{j \neq -1} \theta_{j} \cdot Year_{j=t} \cdot Kids_{i} \cdot Treat_{i} (11) + X_{i}\phi + \vartheta \cdot U_{st} + \lambda \cdot U_{st} \cdot Kids_{i} + \nu_{imt}.$$

This is an extension of equation (2) where we add a dummy variable $Treat_i$ and its interactions with year and kids. This dummy captures the treatment status of a woman in terms of her marital status and spousal earnings. Two definitions of treatment status are considered: one in which $Treat_i = 1$ if a woman is either single or married with spousal earnings of zero (and $Treat_i = 0$ if she is married with positive spousal earnings), and one in which $Treat_i = 1$ if a woman is single (and $Treat_i = 0$ if she is married). In equation (11), the coefficient θ_t is a triple-differences estimate of the effect in year t: it captures the employment change in year t (relative to a base year before the reform) for treated women with vs without kids relative to untreated women with vs without kids. The difference-in-differences estimate for treated women in year t equals $\delta_t + \theta_t$, while the difference-in-differences estimate for untreated women in year t equals δ_t .

F Synthetic Control Study of State Reforms

State EITC reforms are analyzed using a synthetic control approach. For each state with an EITC supplement, a synthetic control state is created from among the pool of states that never had an EITC supplement. Then I run a stacked event study comparing treatment and synthetic control states around state EITC introductions. Table A.2 lists all states with an EITC supplement and provides key details about those supplements.

F.1 Selected State Reforms

There are a total of 27 states that implemented and maintained an EITC supplement for at least 3 years.⁸⁰ I exclude six states from my analysis (Iowa, Maryland, Minnesota, Rhode Island, Vermont and Wisconsin) due to small sample sizes. These states have well below 100 single mothers per year around the time of the reform. I exclude another three states (Indiana, Ohio, and Oregon) because their EITC introduction was offset by other tax changes so that the total ATR was increased. The reform-induced change in the average tax rates can be seen from Table A.2. These tax changes are calculated by setting earnings equal to the first kink of the federal EITC for women with *n* children in year *t*, as described in Section C.

Many state supplements were very small in size, so I also select the ten largest reforms in order to make it easier to detect any positive effect. The ten largest reforms are selected as those reforms which induce a reduction in ATR of at least 3pp as shown in Table A.2. These states are Colorado, Connecticut, District of Columbia, Kansas, Massachusetts, Michigan, Nebraska, New Jersey, New Mexico, and New York.

F.2 Constructing Synthetic Controls

To run the synthetic control analysis, CPS data is collapsed into state-by-year observations. The data is collapsed separately for single women with and without children. I consider an event study window from five years before to six years after each reform. Since the monthly files can only be used from 1989 onwards, to ensure that each reform has a consistent dataset across the

⁸⁰A total of 30 states have instituted a supplement (see Table A.2). But the state of Washington never funded and paid out the credit, while Hawaii and South Carolina instituted their supplements only in 2018.

event window, the analysis uses the March CPS files alone for reforms that occurred before 1993 and the March and monthly CPS files combined for reforms that occurred from 1993 onwards. Event time is set to zero in the first year after the introduction of the EITC supplement.

For each treatment state, a synthetic control state is constructed from states that never had a supplement, matching on the level of the outcome variable (employment or participation rate) in each of the five pre-reform years. Synthetic control regressions are run using the Stata command synth.⁸¹

F.3 Stacked Event Study

I create the stacked event studies using two different approaches. The first is a difference-indifferences approach where the synthetic control regressions are run on the sample of single women with children comparing different states over time. The second is a triple-differences specification which also exploits the variation between those with and without children within states.

Having obtained measures of employment in treatment and synthetic control states, a stacked event study specification is used to estimate the average effect across all state EITC reforms. The event studies shown in Panels A and C of Figure 11 are based on the following specification

$$P_{st} = \sum_{j} \alpha_{j} \cdot Event_{j=t} + \beta \cdot Treat_{s} + \sum_{j \neq -1} \gamma_{j} \cdot Event_{j=t} \cdot Treat_{s} + \nu_{st},$$
(12)

where P_{st} is the employment rate in state *s* at time *t*, $Event_{j=t}$ is an indicator for event time *t*, and $Treat_s$ is an indicator for treatment states. The treatment series (solid line) corresponds to the coefficient $\hat{\alpha}_t + \hat{\gamma}_t$, while the synthetic control series (dashed line) corresponds to $\hat{\alpha}_t$.

Panels B and D of Figure 11 is based on a similar analysis, but it adds the variation from children (within states) to the variation across states in a triple-differences design. Specifically, the analysis is based on the following specification

⁸¹See http://fmwww.bc.edu/RePEc/bocode/s/synth.html for documentation on the synth command.

$$P_{kst} = \sum_{j} \alpha_{j} \cdot Event_{j=t} + \beta \cdot Kids_{k} + \gamma \cdot Treat_{s} + \delta \cdot Kids_{k} \cdot Treat_{s} + \sum_{j \neq -1} \zeta_{j} \cdot Event_{j=t} \cdot Kids_{k} + \sum_{j \neq -1} \eta_{j} \cdot Event_{j=t} \cdot Treat_{s}$$

$$+ \sum_{j \neq -1} \theta_{j} \cdot Event_{j=t} \cdot Kids_{k} \cdot Treat_{s} + \nu_{kst},$$
(13)

where P_{kst} is the employment rate for those with kids status k (with or without) in state s at time t. Here, the treated series (solid line) corresponds to the coefficient $\hat{\zeta}_t + \hat{\theta}_t$, while the synthetic control series (dashed line) corresponds to $\hat{\zeta}_t$.

G Event Studies in the Employment+Welfare Rate

The AFDC/TANF program always served fewer families than the EITC program (see Figure A.1). It is therefore natural to ask if welfare reform could feasibly account for the employment increases in the 1990s, or if the program was too small. The answer to this question depends, not on welfare caseload levels, but on welfare caseload changes. To see if the caseload changes were big enough to explain the effects, I consider a different outcome: the fraction of single women who are either employed or on AFDC/TANF. I refer to this outcome as the "employment+welfare rate".⁸² Given welfare caseloads were falling through the 1990s, the employment+welfare rate will show less of an effect than the employment rate. However, if there is still a positive effect in this outcome, then the caseload drops were too small to account for the entire employment effect, leaving something to explain for non-welfare factors.

The results are shown in Figures A.100-A.101 (for the full sample and by number of children, respectively). These figures compare event studies of the 1993 reform for the employment rate and the employment+welfare rate. While the event study in employment shows large effects in every sample, the event study in employment+welfare is flat or slightly falling in every sample. In other words, once the effect of AFDC/TANF caseload changes is neutralized, there is no remaining treatment effect in the 1990s. This implies that welfare reform *was* large enough to explain the effects. In fact, movements between AFDC/TANF and employment account almost exactly for the employment increases year by year through the 1990s.

⁸²Because single women without children are ineligible for AFDC/TANF, there is no distinction between the employment rate and the employment+welfare rate for this group. For single women with children, to avoid double counting people who are both employed and on welfare, I add only AFDC/TANF participants who are not also employed.

These results are more than just an accounting exercise. This is because EITC and welfare reform are predicted to affect the employment+welfare rate differently. Cutting welfare pushes people from welfare into work or into searching for work. If everyone finds work, the employment+welfare rate will be unaffected, while otherwise it will decrease. Expanding the EITC provides work incentives to people from all non-working states, including those in the AFDC/TANF program, those in other social assistance programs, and those receiving no social assistance. If people respond to the EITC, the employment+welfare rate should increase. The patterns shown here are therefore consistent with welfare reform and inconsistent with EITC reform.

A potential concern with this analysis is that AFDC/TANF participation is underreported in the CPS data and that such underreporting may have changed over time (see Meyer and Mittag 2019; Meyer, Mok, and Sullivan 2015). It is possible to address this concern using publicly available administrative data on welfare caseloads from the Department of Health and Human Services. Because the publicly available data does not provide caseloads by marital status, this analysis has to be done for all women together. An earlier version of this paper provided such an analysis (results available upon request). The results on the employment+welfare rate from CPS data survive in the administrative data: the extensive margin increases in the 1990s disappear when considering the employment+welfare rate outcome.

Appendix References

- BLANK, REBECCA M. (2001): "What Causes Public Assistance Caseloads to Grow?" Journal of Human Resources, 36(1), 85–118.
- BLANK, REBECCA M., AND PATRICIA RUGGLES (1996): "When Do Women Use Aid to Families with Dependent Children and Food Stamps? The Dynamics of Eligibility Versus Participation," *Journal of Human Resources*, 31(1), 57–89.
- COHANY, SHARON, ANNE POLIVKA, AND JENNEFER ROTHGEB (1994): "Revisions in the Current Population Survey Effective January 1994," *Employment & Earnings*, 41.
- CURRIE, JANET (2006): "The Take-up of Social Benefits," in *Public Policy and the Income Distribution*, ed. by Alan J. Auerbach, David Card, and John M. Quigley, pp. 80–148. New York: Russell Sage Foundation.
- FLOOD, SARAH, AND JOSÉ PACAS (2017): "Using the Annual Social and Economic Supplement as Part of a Current Population Survey Panel," *Journal of Economic and Social Measurement*, 42(3–4), 225–248.
- JONES, MAGGIE, AND JAMES ZILIAK (2022): "The Antipoverty Impact of the EITC: New Estimates from Survey and Administrative Tax Records," *National Tax Journal*, 75(3), 451–479.
- MEYER, BRUCE, AND NIKOLAS MITTAG (2019): "Using Linked Survey and Administrative Data to Better Measure Income: Implications for Poverty, Program Effectiveness, and Holes in the Safety Net," *American Economic Journal: Applied Economics*, 11(2), 176–204.
- MEYER, BRUCE, WALLACE MOK, AND JAMES SULLIVAN (2015): "Household Surveys in Crisis," *Journal of Economic Perspectives*, 29(4), 199–226.