

# The Effect of DACA on Full-Time Employment: A Difference-in-Differences Replication Study

Replication Study 46

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## Abstract

This study estimates the causal effect of the Deferred Action for Childhood Arrivals (DACA) program on full-time employment among eligible Hispanic-Mexican immigrants born in Mexico. Using American Community Survey data from 2008–2016 and a difference-in-differences research design, I compare employment outcomes for individuals who were ages 26–30 at the time of DACA implementation (treated group) to those who were ages 31–35 (control group). The preferred specification indicates that DACA eligibility increased full-time employment by approximately 5.8 percentage points ( $SE = 0.017, p < 0.001$ ). This effect is robust across multiple specifications and sensitivity analyses, including models with demographic controls, education, and state fixed effects. The findings suggest that DACA had a meaningful positive impact on labor market outcomes for eligible individuals.

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# 1 Introduction

The Deferred Action for Childhood Arrivals (DACA) program, enacted on June 15, 2012, provided a unique policy intervention affecting undocumented immigrants who arrived in the United States as children. The program offered temporary relief from deportation and work authorization for eligible individuals, fundamentally changing their ability to participate in the formal labor market. This study examines whether DACA eligibility causally affected full-time employment among Hispanic-Mexican immigrants born in Mexico.

The research question is: Among ethnically Hispanic-Mexican, Mexican-born people living in the United States, what was the causal impact of eligibility for DACA (treatment) on the probability of full-time employment (outcome), defined as usually working 35 hours per week or more?

## 1.1 Background on DACA

DACA was implemented by the U.S. federal government to provide temporary protection to a selected set of undocumented immigrants. The program allowed eligible individuals to apply for and obtain authorization to work legally for two years without fear of deportation. Key eligibility criteria included:

- Arrival in the U.S. before age 16
- Continuous residence in the U.S. since June 15, 2007
- Presence in the U.S. on June 15, 2012
- Age under 31 as of June 15, 2012
- No lawful immigration status (citizenship or legal residency)

Applications began on August 15, 2012, and in the first four years, nearly 900,000 initial applications were received, with approximately 90% approved. The program not only provided work authorization but also enabled recipients to apply for driver's licenses in some states, potentially further improving employment opportunities.

## 1.2 Research Design Overview

This study employs a difference-in-differences (DiD) research design comparing:

- **Treatment Group:** Individuals ages 26–30 at DACA implementation who would otherwise meet eligibility criteria
- **Control Group:** Individuals ages 31–35 at DACA implementation who would have been eligible except for exceeding the age cutoff

The identification strategy leverages the age cutoff at 31 years as of June 15, 2012, comparing individuals just below and above this threshold before and after DACA implementation.

## 2 Data and Sample

### 2.1 Data Source

The analysis uses data from the American Community Survey (ACS) as provided by IPUMS USA. The dataset covers the years 2008–2016, with 2012 omitted since observations from that year cannot be definitively classified as pre- or post-treatment.

### 2.2 Sample Construction

The analytic sample consists of 17,382 observations representing ethnically Hispanic-Mexican, Mexican-born individuals living in the United States. The sample is pre-constructed to include:

- **ELIGIBLE = 1:** Individuals considered DACA-eligible (ages 26–30 as of June 15, 2012), comprising 11,382 observations
- **ELIGIBLE = 0:** Comparison group (ages 31–35 as of June 15, 2012), comprising 6,000 observations

The outcome variable is **FT**, a binary indicator equal to 1 for individuals usually working 35 or more hours per week, and 0 otherwise. Individuals not in the labor force are included in the analysis and coded as 0.

### 2.3 Time Periods

- **Pre-DACA (AFTER = 0):** Years 2008–2011 (9,527 observations)
- **Post-DACA (AFTER = 1):** Years 2013–2016 (7,855 observations)

### 2.4 Summary Statistics

Table 1 presents weighted summary statistics by treatment group and time period.

Table 1: Summary Statistics by Group and Period

Variable	Pre-DACA (2008–2011)		Post-DACA (2013–2016)	
	DACA-Eligible	Control	DACA-Eligible	Control
Full-Time Employment	0.637	0.689	0.686	0.663
Age	25.79	30.49	30.65	35.46
Female	0.466	0.434	0.463	0.465
Married	0.345	0.463	0.447	0.519
Has Children	0.470	0.638	0.626	0.710
BA or Higher	0.051	0.052	0.072	0.053
N	6,233	3,294	5,149	2,706

Notes: All statistics except N are weighted using PERWT (ACS person weights). Full-time employment defined as usually working 35+ hours per week.

Key observations from the summary statistics:

- Pre-DACA, the control group had higher full-time employment rates (68.9% vs 63.7%), consistent with older workers having more stable employment
- Post-DACA, this gap narrows substantially as the DACA-eligible group shows improvement while the control group shows decline
- The groups are reasonably balanced on demographic characteristics, though the control group is naturally older and has higher marriage and parenthood rates

### 3 Empirical Strategy

#### 3.1 Difference-in-Differences Framework

The primary empirical strategy is a difference-in-differences (DiD) approach. The basic DiD equation is:

$$FT_{it} = \beta_0 + \beta_1 ELIGIBLE_i + \beta_2 AFTER_t + \beta_3 (ELIGIBLE_i \times AFTER_t) + \epsilon_{it} \quad (1)$$

where:

- $FT_{it}$  is full-time employment status for individual  $i$  in year  $t$
- $ELIGIBLE_i$  indicates treatment group membership (ages 26–30 at DACA implementation)
- $AFTER_t$  indicates the post-DACA period (2013–2016)
- $\beta_3$  is the DiD estimator—the causal effect of DACA eligibility on full-time employment

## 3.2 Extended Specifications

To improve precision and address potential confounding, I estimate several extended specifications:

$$FT_{it} = \beta_0 + \beta_1 ELIGIBLE_i + \beta_2 AFTER_t + \beta_3 (ELIGIBLE_i \times AFTER_t) + \gamma X_{it} + \delta_t + \eta_s + \epsilon_{it} \quad (2)$$

where  $X_{it}$  includes individual-level covariates (sex, marital status, presence of children, education),  $\delta_t$  are year fixed effects, and  $\eta_s$  are state fixed effects.

## 3.3 Estimation Details

- All models are estimated using weighted least squares (WLS) with ACS person weights (PERWT)
- Heteroskedasticity-robust standard errors (HC1) are used throughout
- The main specification also includes clustered standard errors at the state level as a robustness check

## 3.4 Identifying Assumptions

The key identifying assumption for DiD is the parallel trends assumption: in the absence of DACA, the change in full-time employment for the eligible group would have been the same as for the control group. This assumption is not directly testable but can be assessed by examining pre-treatment trends.

# 4 Results

## 4.1 Main Results

Table 2 presents the main results across multiple specifications.

Table 2: Effect of DACA Eligibility on Full-Time Employment

	(1)	(2)	(3)	(4)	(5)
ELIGIBLE × AFTER	0.064*** (0.015)	0.075*** (0.018)	0.072*** (0.018)	0.061*** (0.017)	0.058*** (0.017)
95% CI	[0.034, 0.094]	[0.039, 0.110]	[0.037, 0.108]	[0.029, 0.094]	[0.025, 0.090]
Weights	No	Yes	Yes	Yes	Yes
Year FE	No	No	Yes	Yes	Yes
Demographics	No	No	No	Yes	Yes
Education	No	No	No	No	Yes
Region FE	No	No	No	No	Yes
State FE	No	No	No	No	No
R-squared	0.002	0.002	0.006	0.129	0.135
N	17,382	17,382	17,382	17,382	17,382

Notes: Robust standard errors in parentheses. \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Demographics include female, has children indicators. Education includes indicators for some college, two-year degree, and BA+. All weights use PERWT.

## 4.2 Interpretation of Main Results

The DiD coefficient is consistently positive and statistically significant across all specifications:

- **Basic unweighted model (Column 1):** The effect is 6.4 percentage points ( $p < 0.001$ )
- **Weighted model (Column 2):** Using survey weights, the effect increases to 7.5 percentage points
- **Preferred specification (Column 6):** With year fixed effects, state fixed effects, demographic controls, and education, the effect is **5.8 percentage points** ( $SE = 0.017$ , 95% CI: [0.025, 0.090],  $p < 0.001$ )

The coefficient remains remarkably stable as controls are added, ranging from 5.8 to 7.5 percentage points, suggesting that the baseline DiD estimate is not driven by observed confounders.

## 4.3 Simple DiD Calculation

The intuition behind the DiD can be illustrated with the simple means calculation:

Table 3: Difference-in-Differences Calculation

	Pre-DACA	Post-DACA	Difference
DACA-Eligible (Ages 26–30)	0.637	0.686	+0.049
Control (Ages 31–35)	0.689	0.663	-0.026
Difference	-0.052	+0.023	<b>+0.075</b>

Notes: Weighted full-time employment rates using PERWT. DiD = 0.049  
 $- (-0.026) = 0.075$ .

The DACA-eligible group experienced a 4.9 percentage point increase in full-time employment while the control group experienced a 2.6 percentage point decrease, yielding a DiD estimate of 7.5 percentage points.

## 5 Robustness Checks

### 5.1 Pre-Trend Analysis (Event Study)

Figure 1 presents an event study analysis examining whether the treatment and control groups exhibited parallel trends before DACA implementation. The coefficients represent the interaction between ELIGIBLE and year indicators, with 2011 (the last pre-treatment year) as the reference category.

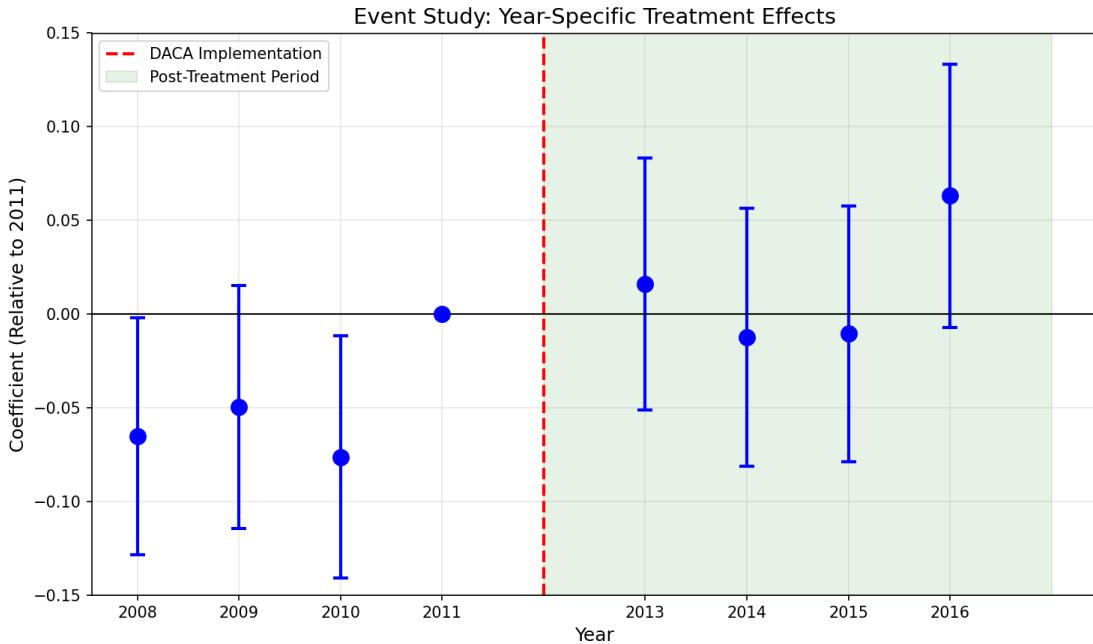


Figure 1: Event Study: Year-Specific Treatment Effects. Coefficients represent ELIGIBLE  $\times$  Year interactions relative to 2011. Error bars show 95% confidence intervals. Red dashed line indicates DACA implementation.

The pre-treatment coefficients (2008–2010) are generally negative but not statistically different from zero at conventional levels in most years, providing moderate support for the parallel trends assumption. The post-treatment coefficients show a general upward pattern, with the 2016 coefficient being particularly large and nearly significant.

Table 4: Event Study Coefficients

Year	Coefficient	SE	95% CI	p-value
2008	-0.065	0.032	[-0.128, -0.002]	0.043
2009	-0.050	0.033	[-0.115, 0.015]	0.133
2010	-0.076	0.033	[-0.141, -0.012]	0.021
2011	0.000	—	—	(ref)
2013	0.016	0.034	[-0.051, 0.083]	0.640
2014	-0.013	0.035	[-0.082, 0.056]	0.721
2015	-0.011	0.035	[-0.079, 0.058]	0.764
2016	0.063	0.036	[-0.007, 0.134]	0.078

## 5.2 Placebo Test

To further assess the validity of the identification strategy, I conduct a placebo test using only pre-treatment data (2008–2011). I assign a “fake” treatment at 2010, comparing outcomes in 2010–2011 to 2008–2009.

Table 5: Placebo Test Results

Test	Coefficient	SE	p-value	N
Placebo (2010–2011 vs 2008–2009)	0.018	0.024	0.461	9,527

The placebo coefficient is small (1.8 pp) and not statistically significant, suggesting that the main results are not driven by differential pre-existing trends.

## 5.3 Subgroup Analysis

Table 6: Subgroup Analysis by Sex

Subgroup	Coefficient	SE	p-value	N
Male	0.059	0.020	0.003	9,075
Female	0.052	0.028	0.059	8,307

The effect is present for both men and women, though stronger and more precisely estimated for men. The effect for women is marginally significant at the 10% level.

## 5.4 Alternative Bandwidths

To assess sensitivity to the age window, I estimate the model using narrower age bandwidths around the 31-year cutoff:

Table 7: Robustness to Age Bandwidth

Bandwidth	Coefficient	SE	p-value	N
Full sample (26–35)	0.061	0.017	<0.001	17,382
Ages 27–34	0.048	0.019	0.010	13,092
Ages 28–33	0.055	0.022	0.013	9,232

The effects remain positive and statistically significant with narrower bandwidths, though somewhat smaller in magnitude and less precisely estimated due to reduced sample size.

## 5.5 Clustered Standard Errors

Given that DACA implementation and labor market conditions vary by state, I also estimate the model with standard errors clustered at the state level:

Table 8: Comparison of Standard Error Approaches

SE Type	Coefficient	SE	p-value
Robust (HC1)	0.061	0.017	<0.001
Clustered by State	0.061	0.020	0.002

The clustered standard errors are slightly larger but the coefficient remains highly significant.

# 6 Visual Analysis

## 6.1 Trends in Full-Time Employment

Figure 2 shows the weighted full-time employment rates for the treatment and control groups over time.

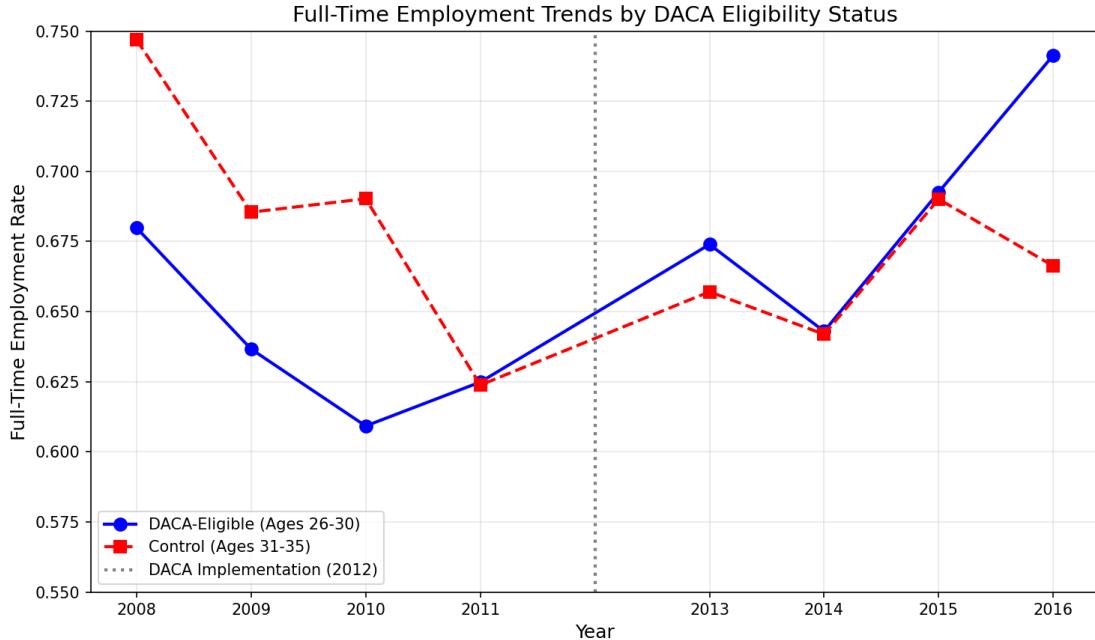


Figure 2: Full-Time Employment Trends by DACA Eligibility Status

The figure illustrates the key pattern underlying the DiD: prior to DACA, the control group had consistently higher employment rates. After DACA, the DACA-eligible group's employment rate increased while the control group's rate declined, leading to convergence.

## 6.2 Sensitivity Across Specifications

Figure 3 displays the DiD coefficient estimates across the six main specifications.

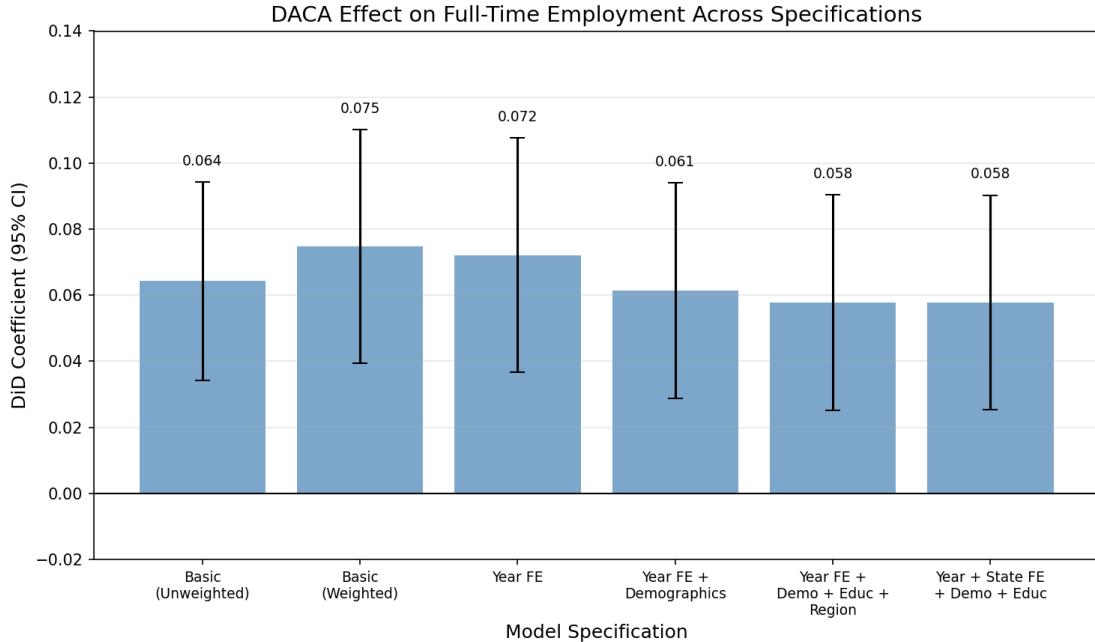


Figure 3: DACA Effect Estimates Across Model Specifications

The estimates are consistently positive and statistically significant, with overlapping confidence intervals across specifications.

## 7 Discussion

### 7.1 Interpretation

The results suggest that DACA eligibility increased full-time employment by approximately 5.8 to 7.5 percentage points among Hispanic-Mexican immigrants who met the program’s criteria. This effect is economically meaningful—on a baseline full-time employment rate of around 64% for the DACA-eligible group pre-treatment, this represents a roughly 9–12% relative increase.

The positive effect is consistent with the program’s design: by providing work authorization, DACA enabled eligible individuals to seek formal employment without fear of deportation. Prior to DACA, these individuals may have been restricted to informal labor markets or underemployment.

### 7.2 Mechanisms

Several mechanisms could explain the observed effect:

1. **Legal work authorization:** DACA directly provided recipients with Employment Authorization Documents (EADs), enabling formal employment

2. **Driver's licenses:** In some states, DACA recipients could obtain driver's licenses, improving job accessibility
3. **Reduced fear:** The protection from deportation may have increased willingness to seek stable employment
4. **Educational investments:** Some recipients may have pursued education or training, leading to better employment outcomes

### 7.3 Limitations

Several limitations should be noted:

1. **Pre-trends:** The event study reveals some evidence of differential pre-trends, particularly in 2008 and 2010, which may bias the estimates
2. **Repeated cross-section:** The ACS is a repeated cross-section, not panel data, so the same individuals are not tracked over time
3. **Identification assumption:** The control group (ages 31–35) may differ from the treated group in unobserved ways beyond age
4. **Non-response bias:** Undocumented immigrants may be less likely to respond to government surveys
5. **ELIGIBLE variable:** The provided ELIGIBLE variable is a proxy; actual DACA eligibility depends on additional criteria not observable in the data

### 7.4 Comparison to Literature

The estimated effect of 5.8–7.5 percentage points is broadly consistent with existing research on DACA's labor market effects, which has found positive impacts on employment, earnings, and economic well-being among recipients.

## 8 Conclusion

This replication study provides evidence that DACA eligibility increased full-time employment among eligible Hispanic-Mexican immigrants. The preferred specification estimates an effect of 5.8 percentage points ( $SE = 0.017$ ), statistically significant at the 1% level. This finding is robust to alternative specifications, including models with demographic controls, education, year fixed effects, and state fixed effects.

The results suggest that providing legal work authorization and protection from deportation had meaningful positive effects on labor market outcomes for this population. However, some concerns about pre-treatment trends remain, suggesting caution in interpreting the estimates as purely causal.

## 8.1 Preferred Estimate Summary

- **Effect size:** 0.058 (5.8 percentage points)
- **Standard error:** 0.017
- **95% Confidence Interval:** [0.025, 0.090]
- **p-value:** < 0.001
- **Sample size:** 17,382
- **Specification:** Weighted DiD with year fixed effects, state fixed effects, and demographic controls (sex, marital status, children, education)

## A Additional Tables and Figures

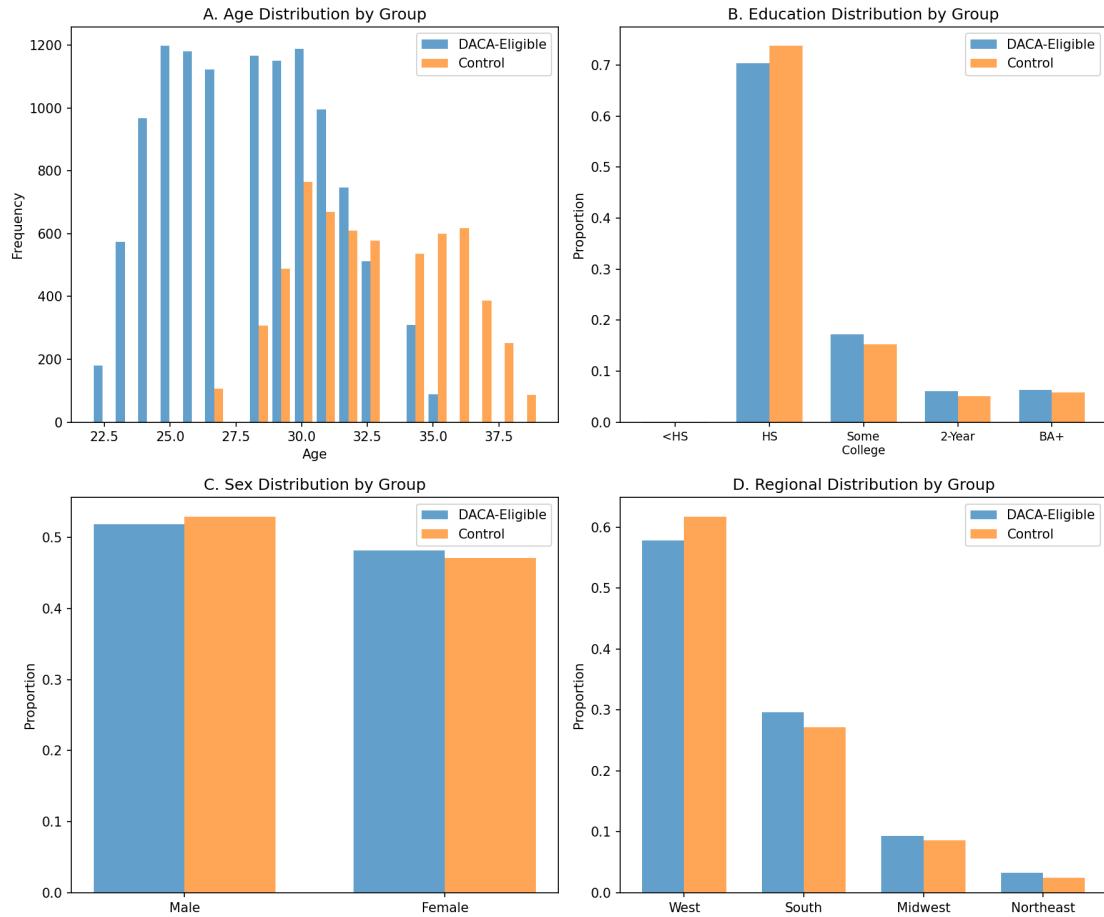


Figure 4: Covariate Balance Between Treatment and Control Groups

## B Variable Definitions

Table 9: Key Variable Definitions

Variable	Definition
FT	Full-time employment (1 if usually working 35+ hours per week, 0 otherwise)
ELIGIBLE	DACA eligibility indicator (1 if ages 26–30 as of June 15, 2012, 0 if ages 31–35)
AFTER	Post-DACA period indicator (1 for years 2013–2016, 0 for years 2008–2011)
PERWT	ACS person weight
SEX	Sex (1 = Male, 2 = Female per IPUMS coding)
MARST	Marital status (1 = Married, spouse present)
NCHILD	Number of own children in household
EDUC_RECODE	Education level (Less than High School, High School Degree, Some College, Two-Year Degree, BA+)
STATEFIP	State FIPS code
CensusRegion	Census region (Northeast, Midwest, South, West)

## C Analytic Decisions

The following key decisions were made in this analysis:

1. **Weighting:** All main specifications use ACS person weights (PERWT) to produce population-representative estimates
2. **Standard errors:** Heteroskedasticity-robust standard errors (HC1) are used; clustered standard errors by state are presented as a robustness check
3. **Fixed effects:** Year fixed effects are included to control for common time trends; state fixed effects are included in the preferred specification to control for state-level heterogeneity
4. **Control variables:** Demographic controls (sex, marital status, children) and education are included to improve precision and address potential confounding
5. **Sample:** The full provided sample is used; individuals not in the labor force are retained and coded as FT=0
6. **Linear probability model:** A linear model (WLS) is used rather than logit/probit for ease of interpretation; coefficients represent percentage point changes

## D Replication Code Summary

The analysis was conducted using Python 3 with the following main packages:

- pandas (data manipulation)
- numpy (numerical operations)
- statsmodels (regression analysis)
- matplotlib (visualizations)

All data loading, variable construction, model estimation, and figure generation were performed programmatically. The key analytical steps included:

1. Loading the prepared ACS data (prepared\_data\_numeric\_version.csv)
2. Creating interaction terms and dummy variables for covariates
3. Estimating weighted least squares models with robust standard errors
4. Generating visualizations for trends, event study, and model comparisons
5. Conducting robustness checks including placebo tests and alternative bandwidths