

THE SURPRISING IMPACTS OF UNIONIZATION ON ESTABLISHMENTS: ACCOUNTING FOR SELECTION IN CLOSE UNION REPRESENTATION ELECTIONS*

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Abstract

Regression discontinuity studies of the effects of union representation rely on quasi-random variation in vote tallies near the majority-rule threshold. It is precisely in the closest elections, however, that incentives may be most powerful for manipulation. This study shows evidence for such contaminating selection in close union elections. First, unions lost close elections at an unexpectedly high rate, even those involving a large number of voters. Second, prior to the election, establishments where the union lost by a narrow margin employed fewer workers, had a smaller payroll, and had a more dispersed distribution of workers earnings than establishments where the union barely won. As a result, level comparisons between establishments where the union barely won and barely lost are misleading. Regression discontinuity panel estimates which account for pre-election differences suggest that union representation had a negative effect on payroll, average worker earnings, and employment, in marked contrast to previous findings. This effect operated largely through compositional changes: higher-paid workers were more likely to leave, while younger and lower-paid workers were more likely to come following a union victory. The union effect on the earnings of workers who stayed was minimal.

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1 Introduction

Close union certification elections offer the hope of transparently identifying the causal effects of union representation. If the distribution of underlying worker and establishment characteristics varies continuously with the union vote share in the neighborhood of the majority-rule threshold, then comparisons of outcomes where the union barely won or barely lost reflect the causal effect of union representation, akin to a randomized controlled experiment (Hahn, Todd, and van der Klaauw, 2001; DiNardo and Lee, 2004; Lee, 2008; Lee and Mas, 2012).

The discontinuity at the majority-rule threshold is a two-edged sword, however. The idea that very small differences in vote shares can lead to large differences in union status and possibly outcomes provides the essence of the identification strategy but also generates powerful incentives for manipulation exactly at the threshold that would invalidate the identifying assumptions. It is precisely the closest elections that are possibly the most susceptible to either *ex ante* or *ex post* manipulation of voting outcomes. If the tendency to manipulate elections is related to the potential outcomes, then comparisons of establishments and workers where the union barely won or barely lost will no longer reflect the causal effect of union representation.

This study shows evidence for such contaminating selection in close union representation elections. First, unions lost close elections at an unexpectedly high rate, even those involving a large number of voters. Employers appeared to be more successful than unions at challenging and throwing out ballots when the election came down to a single vote. Second, using worker-level and establishment-level panel data from the U.S. Census Bureau, I find that prior to the election, establishments where the union lost by a narrow margin employed fewer workers, had a smaller payroll, and had a more dispersed distribution of workers earnings than establishments where the union barely won. As a result, level comparisons between establishments where the union barely won and barely lost are misleading.

Accounting for pre-election differences gives surprising results. Regression discontinuity estimates that exploit the panel nature of the data to control for underlying differences suggest that union representation had a substantial negative effect on payroll, average worker earnings, and, to a lesser extent, employment, in marked contrast to previous findings. These effects operated largely through compositional changes in an establishment's workforce in response to a union vic-

tory: higher paid workers were more likely to leave the establishment, while younger and lower-paid workers were more likely to come following a union victory. The union impact on the earnings of workers who stayed was minimal.

The methods and findings in this paper relate to two strands of the long literature on union impacts and reconcile some conflicting results. One strand has attempted to measure unionization's impact on establishment outcomes. Early studies in this strand used a difference-in-differences design comparing establishments before and after unionizing drives, and found that unionization reduced the scale of production at unionizing establishments with little effect on workers' wages (Freeman and Kleiner, 1990; Lalonde, Marschke, and Troske, 1996). Later studies in this same vein, however, using regression discontinuity designs comparing post-election outcomes of establishments where unions barely won or barely lost found little difference in any establishment outcomes, including the scale of production (DiNardo and Lee, 2004; Lee and Mas, 2012). This study combines the two methods and reconciles the contrasting findings: post-election employment and payroll were indeed similar where the union barely won or lost, but evidence which takes into account pre-election selection suggests that unionization led to significant decreases in the scale of production as measured by employment and payroll, but also reduced average worker earnings at the establishment.

A question left open by these establishment-level studies is what impacts unionization has on individual workers, a topic addressed by the extensive literature on individual union wage premia. This literature was summarized by Freeman and Medoff (1984) and more recently by Blanchflower and Bryson (2003), and has found consistently large positive wage gaps between observably comparable union and nonunion workers, in contrast to the small or even negative establishment-level estimates. Even apart from differences in sample and research design, a discrepancy between establishment- and individual-level union effects need not be a contradiction, however. Effects on worker composition or distributional effects may mask underlying effects on individual workers. This paper shows that compositional and distributional effects were indeed driving the lion's share of the establishment-level effects, although it rejects large, positive individual effects.

This paper also relates to the methodological literature on RD designs in general and election-based studies in particular. First, the findings underscore the need for pre-treatment "balance" tests as suggested by Imbens and Lemieux (2008) and Lee and Lemieux (2010). Especially where

the discontinuity’s stakes are high for the agents involved, the burden should be on the researcher to provide powerful tests that manipulation or sorting is not present at the threshold. For example, not only the levels but also the distribution of lagged outcomes should be balanced across the threshold. Second, this study adds to existing evidence that close elections, even those involving a large electorate, exhibit substantial selection (Caughey and Sekhon, 2011). RD results based on close elections should therefore be interpreted with caution.

Methodologically, evidence for selection near the threshold does not necessarily sink the possibility of valid causal inference. Panel data—often used to detect selection in the first place—can be used to construct estimators that account for confounding differences across the threshold, albeit only under certain conditions. This paper estimates standard RD local linear nonparametric specifications in first differences in order to control for pre-election selection. The identifying assumption is that the factors underlying selection in close elections are time-constant. This type of identification strategy is a hybrid of RD and a traditional difference-in-differences research design.

The next section describes the institutional details of U.S. private sector union representation elections. Section 3 describes the data used in the empirical work. Section 4 lays out the econometric framework for identifying and estimating effects in the regression discontinuity design in the presence of confounding selection. Section 5 presents evidence for non-random selection in private sector union representation elections as well as estimates of the effects of union representation which account for the selection. Section 6 summarizes the findings and concludes.

2 Background

Since 1935, most U.S. private sector unionization has been governed by the National Labor Relations Act (NLRA), which specifies the rights of unionizing workers. The traditional process by which workers unionize is through a National Labor Relations Board (NLRB) secret ballot election, although an employer may voluntarily bargain with the workers’ chosen representative, or in some cases may be required to do so even without an election.¹ The following steps describe the nominal path a group of workers follows to form a union:²

¹Secret ballot election has historically been the dominant form of new unionization, although in recent years voluntary recognition through neutrality agreements and card checks have become more common. (Brudney, 2005)

²The simple process laid out here follows the procedures described in NLRB (2010).

1. Petition drive: Union organizers lobby workers, collect signatures expressing a desire for union representation, and submit a petition to the NLRB to hold an election. If the petition is accepted, the NLRB ascertains the scope of the bargaining unit and sets the election time and place, usually the workplace.
2. Election: Eligible workers vote for or against the union, and the union wins if it receives a strict majority of the votes cast.
3. Certification: If the union wins, the NLRB certifies it as the sole authorized representative of the workers in the bargaining unit, and requires the employer to bargain “in good faith” with the union.

In practice, however, the union certification process can be fraught with disputes and delays, before and after the election (Ferguson, 2008). For example, the employer or union may challenge the scope of the bargaining unit on the initial petition, prior to the election. Any disputes on the bargaining unit scope are resolved by the NLRB prior to the election, after hearings involving both parties, if necessary. During the election itself and subsequent vote count, either side may challenge the validity of individual ballots, and if the challenged ballots could be pivotal in the election, the NLRB rules on whether to count them or throw them out. The employer or union may submit charges of unfair labor practices at any stage, before or after the election. Unfair labor practices are actions on the part of the employer or the union that are deemed to violate the required “laboratory conditions” of a secret-ballot election, and if the charges are found by the NLRB to be of merit, the results of the election are set aside (Moberly, 2002). Unionizing drive disputes are not rare: charges of unfair labor practices were filed in over 21 percent of representation cases between 1999 and 2004 (Ferguson, 2008).

Disputes like these can lead to interventions in the nominal election process that introduce non-random selection even in close elections involving a large number of voters. Interventions after the election that change the voting outcome—for example, enough ballots are successfully challenged and thrown out that the outcome is reversed—can obviously introduce selection. Charges of unfair labor practices may be strategically leveled after a close election loss, also leading to selection in close elections. However, even pre-election interventions can introduce selection if the employer or union can accurately forecast the potential results of an election. More subtly, an unresolved

pre-election dispute can lead to ex-post selection when the supervising authority decides to take up the matter after the election only if it may have been determinative. Disputes are, of course, most likely to be determinative if the election is close.

3 Data

3.1 Union Elections

The analysis uses a dataset on the universe of NLRB union representation election results from 1980 to 2009, combining data obtained from Hank Farber, J.P. Ferguson, and Thomas Holmes.³ Each record in this dataset represents a union certification election held at an establishment, and includes the number of votes cast for and against union representation, the date of the election, and the employer's name and address. The employer name and address information was used to match to Census establishment and individual earnings data (see below).

Table I reports statistics on election characteristics and outcomes for this sample. The sample includes 45,176 representation elections, involving over four million votes cast. The average number of voters participating in these elections was 93, and the union won 45 percent of them. The average union vote share was 49 percent.

3.2 Payroll and Employment

Establishment-level data on payroll and employment come from the Census Bureau's Longitudinal Business Database (LBD). The LBD includes yearly data on nearly all non-farm private sector employers from 1976 to 2009.⁴ An advantage of this database over the Longitudinal Research Database (LRD) used by DiNardo and Lee (2004) is that the LRD is restricted to establishments in the manufacturing sector. Establishments in the LBD are defined as a single physical location where business is conducted. Thus, two places of business owned by the same enterprise are distinct establishments in the LBD. Over 23 million establishments are included in the LBD, although this study focuses only on those where a union certification election was held from 1980 to 2009.

Union election establishments were identified by in the LBD by matching on employer name

³The union data from Thomas Holmes is available on his website, http://www.econ.umn.edu/~holmes/data/geo_spill/

⁴The description of the LBD in this section is based on Jarmin and Miranda (2002).

and address information. The matching was performed by first standardizing the name and address fields in both datasets using an automated procedure and merging on the standardized values. See the appendix for more details on the matching algorithm. This procedure succeeded in identifying 82 percent of the elections in the NLRB dataset.⁵

The LBD includes yearly payroll and employment data for each establishment, derived from administrative payroll tax data from the Internal Revenue Service (IRS). Table II reports statistics on establishment characteristics from the LBD sample. Establishments where the union won the election had larger employment and payroll than where the union lost, both prior to and after the election. Post-election average worker earnings were lower at plants where the union won, however.

3.3 Individual Earnings

Individual-level earnings were obtained from the U.S. Census Bureau’s Longitudinal Employer-Household Dynamics (LEHD) database. The LEHD integrates the universe of unemployment insurance-covered (UI) earnings records held by participating state agencies into a cohesive data structure using person and employer identifiers, allowing linkages to other sources of data.⁶

The Employment History Files (EHF) within the LEHD contain quarterly records of individuals’ UI-covered earnings. The EHF for each of the 30 covered states contains a record for each employee-employer combination—a job—that produced at least one dollar of wages in that state in each year. The data cover a period as wide as 1985 to 2008, although for most states the data only go back to the early 1990s. The EHF contains more than 2.8 billion records, although I focus on workers employed at establishments where a union representation election was held.

Workers employed at election establishments were identified using the NLRB-LBD match as a starting point and merging on establishment identifiers in the Census Bureau’s Business Register Bridge (BRB), which links establishments in the LBD with the LEHD. The overall procedure matched 77 percent of the NLRB elections held in states and years where the LEHD is available, identifying over 1.7 million individuals employed at election establishments.

Table III reports statistics on worker characteristics and earnings from the matched LEHD

⁵For comparison, the match rate in DiNardo and Lee (2004) was 26 percent.

⁶For more details on the construction and uses of the LEHD database, see McKinney and Vilhuber (2008), Lane (2008), Abowd, Haltiwanger, and Lane (2004), and Abowd, Stephens, Vilhuber, Andersson, McKinney, Roemer, and Woodcock (2009).

sample. Employees at establishments where the union won were slightly older, and slightly more likely to be female and nonwhite. The percentage of workers at establishments before a union election who remained at the same establishment following the election was slightly more than 40 percent.⁷

4 Econometric Framework

4.1 Identification

A fundamental obstacle to measuring the effect of unionization is selection bias: outcomes within unionized plants may differ for reasons other than union representation. One promising approach to overcoming these selection issues, first used in this context by DiNardo and Lee (2004), is a regression discontinuity design based on close union representation elections. If plants and workers where the union barely won and barely lost are comparable, then close union elections approximate a randomized experiment, and the resulting difference in the distribution of outcomes provides a reliable estimate of the causal effect of unionization.

To formalize this idea, let $D = 1(R > 0.5)$ be an indicator for a union victory, where R is the union vote share. Let Y_1 be the outcome that would be realized if the union were to win, and let Y_0 be the outcome otherwise, so that the observed outcome is $Y = Y_0 + (Y_1 - Y_0)D$. The impacts of unionization are captured by comparisons of the conditional distributions of Y_0 and Y_1 at the threshold of the running variable, $R = 0.5$. Naturally, if the impacts of union representation are heterogeneous, this is a local effect specific to units at the threshold.

The key assumption for identifying the impact of union representation in this setting is that the conditional distribution of potential outcomes as a function of the union vote share is continuous near the threshold of union victory, and thus any jumps in the observed outcome distribution at the threshold are due to the treatment. Formally:

Assumption 1: Local Continuity $F_{Y_d|R}(y|r)$ is continuous in r at $r = 0.5$, for $d \in \{0, 1\}$.

This assumption is likely to be satisfied if, for example, unions, workers, and firms are *a priori* uncertain about the outcome of the election when it is close (see Lee, 2008 for a formal proof).

⁷This would seem low given typical worker turnover rates; however, note that the denominator includes matched workers at any time prior to the election, including those who may have left before the election.

The requisite uncertainty may in turn be more plausible for elections involving a relatively large number of voters, which is why studies using this design often restrict to elections involving 20 or more votes (DiNardo and Lee, 2004; Lee and Mas, 2012; Sojourner, Town, Grabowski, Chen, and Frandsen, 2013).

This assumption will be violated, however, if employers or unions can (with positive probability) accurately forecast the voting outcome and take action to influence it, or alter the election results ex post. The institutional mechanisms of ballot challenges and unfair labor practice accusations are plausible channels through which such ex post contamination could occur.

It is far from certain a priori, therefore, that the RD continuity assumption holds in the union election setting. Fortunately, the RD design lends itself to powerful specification tests. Since Y_0 and Y_1 are only observed on one side of the threshold or the other, the continuity assumption cannot be directly tested, but plausible rationales for Assumption 1 will also imply that the density of R and the conditional distributions of pre-treatment characteristics, including lagged outcomes, will be continuous at $R = 0.5$. These implications can be tested directly and, depending on their power, can provide compelling evidence for or against Assumption 1 (McCrary, 2008; Imbens and Lemieux, 2008; Lee and Lemieux, 2010).

If tests show evidence of discontinuities in the conditional distribution of pre-election characteristics at the threshold, then Assumption 1 is questionable. If pre-election outcomes, Y_{-1} , are observed, then an alternative identifying assumption that relaxes Assumption 1 can be made. Define $\Delta_d \equiv Y_d - Y_{-1}$ to be the potential change in the outcome following a union election, with its conditional cdf given by $F_{\Delta_d|R}(y|r)$. The alternative identifying assumption imposes continuity on the conditional distribution of Δ_d :

Assumption A1': Local Continuity in Differences $F_{\Delta_d|R}(y|r)$ is continuous in r at $r = 0.5$, for $d \in \{0, 1\}$.

If the conditional distribution of Y_{-1} is continuous at the threshold, then assumption A1 implies A1'. However, A1' may hold even if A1 is violated, if the factors underlying the selection have a time-invariant impact on outcomes.

If the traditional RD assumption A1 holds, then the average treatment effect at the threshold is identified by comparing the conditional means of post-election outcomes just above and below

the threshold: $\delta = \lim_{r \rightarrow 0.5^+} E[Y|R=r] - \lim_{r \rightarrow 0.5^-} E[Y|R=r]$. If the weaker assumption A1' holds, then the effect is identified by comparing the conditional means of post-election first differences:

$$\delta = \lim_{r \rightarrow 0.5^+} E[Y - Y_{-1}|R=r] - \lim_{r \rightarrow 0.5^-} E[Y - Y_{-1}|R=r].$$

4.2 Estimation and Testing

Following established RD methodology, estimation and testing will consist of graphical analysis accompanied by statistical inference based on local polynomial regressions (Lee and Lemieux, 2010; Porter, 2003). The prototypical RD graphs plot averages of the dependent variable by non-overlapping bins of the running variable that do not straddle the threshold. Local polynomial regressions complement the graphical analysis by allowing formal hypothesis tests and confidence intervals. A typical local linear specification uses weighted least squares to estimate

$$Y_i = \beta_0 + \delta D_i + \beta_1 R_i + \beta_2 R_i D_i + \varepsilon_i$$

with weights $w_i = K\left(\frac{R_i}{h}\right)$, where K is a kernel function and h is a bandwidth chosen by, say, the procedure in Imbens and Kalyanaraman (2012). The estimate of δ in this specification is the local linear estimator for the average treatment effect on Y at the threshold. If the dependent variable is a pre-treatment characteristic such as a lagged outcome, then a test of $H_0 : \delta = 0$ is a specification check for assumption A1. Testing the continuity of the density of R at the threshold can be carried out as described in McCrary (2008).

If tests of A1 fail, a specification which relies on the weaker assumption A1' and accounts for time-invariant selection at the threshold is

$$Y_i - Y_{i,-1} = \beta_0 + \delta D_i + \beta_1 R_i + \beta_2 R_i D_i + \varepsilon_i, \tag{1}$$

where $Y_{i,-1}$ is a pre-election observation and local linear weighting is also used. This specification is a local linear difference-in-differences specification where the treatment effect δ is identified as the difference between the before-after comparisons across the threshold.

The union margin of victory, R_i , will be adjusted as in DiNardo and Lee (2004) to avoid mechanical discontinuities stemming from differences in the support of the union vote share for

elections of different sizes. First, only elections with k or more voters will be included. Second, the raw vote shares will be binned in $100/k$ -percent intervals. The support of this adjusted vote share will thus be identical for all included elections. To be precise, if U_i is the number of votes for the union and T_i is the total number of votes cast, the running variable will be constructed as:

$$R_i = (\lceil kU_i/T_i \rceil - 0.5) / k,$$

where $\lceil \cdot \rceil$ is the ceiling function. Following Dinardo and Lee k will initially be set equal to 20, but in specification tests will vary over a range.

Finally, all analyses will be performed after collapsing the data on values of the adjusted running variable, and weighted by the number of establishments or individuals. The collapsed data satisfy Census Bureau confidential disclosure requirements and, of course, yield identical point estimates to regressions on the original micro data. Heteroskedasticity-consistent inference using the collapsed data is asymptotically equivalent to clustering on the values of the adjusted running variable, as suggested by Lee and Card (2008) to account for specification error arising from a discrete running variable.

5 Results

5.1 Non-random selection in close elections

5.1.1 Vote share density discontinuities

The density of the union vote share shows strong evidence of sorting near the threshold. Figure I shows the distribution of the union vote share (the number of votes for the union divided by the total number of votes) in the sample in 5-percentage point bins. In this figure and elsewhere the analysis is restricted to elections where at least 20 votes were cast, unless otherwise indicated. The mode is around 40 percent, with a significant number of elections in which the union received all votes. The frequency of the bin corresponding to the closest union victories appears to be abnormally low relative to surrounding bins, and a McCrary (2008) test strongly rejects continuity at the threshold (t-stat ≈ 10). The plot suggests the presence of some manipulation or selection which resulted in “too few” establishments experiencing a close union victory.

Sorting in close elections was substantial even in elections with a large number of voters. Previous studies using union elections restricted to elections with at least 20 voters because elections with a larger number of voters were presumably less susceptible to manipulation. However, the union vote share density discontinuity remains large even when restricting to elections with a large number of voters. Figure II plots the density of the union vote share restricting to elections with at least $k \in \{20, 30, 40, 60, 80, 100\}$ voters. The upper-left panel reproduces Figure I with a cutoff of 20 voters. The “hole” in the density corresponding to close union victories persists as larger and larger cutoffs are chosen. The discontinuity is clearly visible even in the lower-right panel which restricts to elections with at least 100 voters. The McCrary density test strongly rejects continuity for all size cutoffs except for 100 where the test becomes less powerful as fewer and fewer elections remain above the cutoff. This evidence suggests that restricting analysis to large elections will not eliminate the selection threat.

The density of the union margin of victory in terms of number of votes also shows strong evidence of sorting and suggests that the mechanism is ex post challenges to the voting results. Figure III shows the distribution of the union margin of victory, defined as the number of union votes minus the number of votes the union needed for victory. The figure shows an anomalous drop in the density at a zero margin of victory, that is, where the union ended up with exactly the number of votes needed to win. This pattern of “too few” close union victories is identical to the pattern of selection seen in terms of the union vote share, and implies that the “missing” union victories are the ones that would have come down to a single vote.

A likely mechanism is that employers were more successful than unions at challenging and throwing out ballots when the election came down to a single vote. The employer would have the strongest incentive to exercise this ability in elections with an odd number of total voters where the union obtained one more vote than the employer, since a single successful challenge would reverse the outcome.⁸ Figure IV plots the distribution of the union margin of victory for elections with an odd number of voters. The figure shows an even more pronounced drop in the density at a zero

⁸The closest margin by which a union can win an odd election would give the union one more vote than the employer. To reverse the outcome of a close odd election, the employer would need to have only one ballot successfully challenged and thrown out. The closest margin by which a union can win an even election, on the other hand, would give the union two more votes than the employer. To reverse the outcome of a close even election, the employer would need to have two ballots successfully challenged. On the other hand, the closest margin by which a union can lose an even election is to tie with the employer. Now, it is the union which only needs to have one ballot successfully challenged and thrown out to reverse the outcome of the election.

margin of victory (that is, where the union won by one vote), consistent with employers successfully challenging a pro-union ballot where doing so would reverse the election’s outcome in their favor. In elections with an even number of voters the situation reverses: only in elections that would have ended in a tie (and hence have an even number of voters) could a single ballot challenge on a part of the *union* reverse the outcome of the election, since by rule a tie favors the employer. Thus for elections with an even number of voters the incentive for manipulation is strongest just *below* the threshold of union victory. Figure V plots the distribution of the union margin of victory for elections with an even number of voters, and supports this hypothesis. The figure shows the opposite pattern: there are unexpectedly few elections where the union tied with the employer, consistent with the union successfully challenging an anti-union ballot in the case of a would-be tie.

The asymmetry in the sorting based on whether the election involved an even or odd number of voters, while at first blush surprising, suggests that the mechanism behind the sorting in vote tallies is the ability to challenge and throw out opposing ballots. Employers appear to be able to exercise this ability more frequently, since the density drop in Figure IV is larger than in Figure V, and the overall sorting in Figures I and III favors the employer.

5.1.2 Discontinuity in pre-election characteristics

If establishments where the employer is most likely able to reverse the election outcome in its favor differ from establishments where the employer can’t (and similarly for the union) then the sorting documented in the previous section will lead to discontinuities in pre-election characteristics at the threshold. This section shows evidence for just such discontinuities in pre-election employment, payroll, and worker earnings.

Establishments where the union barely won employed significantly more workers prior to the election than establishments where the union barely lost. Figure VI plots average log pre-election employment from the LBD by non-overlapping 5-percent bins in the union vote share. There is a substantial discontinuity at the threshold of union victory.⁹ Local linear estimates in the top row of Table IV show that establishments where the union barely won employed about .13 (s.e. = .032) log points more workers prior to the election than establishments where the union barely lost.

⁹In Figure VI and others there are jumps away from the threshold as well. Much of this is noise: as Figures I and II show, the number of elections drops off quickly at vote shares away from the threshold.

Likewise, establishments where the union barely won had larger payrolls prior to the election than establishments where the union barely lost. Figure VII plots average log pre-election payroll from the LBD by non-overlapping 5-percent bins in the union vote share. Establishments where the union barely won have an unexpectedly high payroll. Local linear estimates in the second row Table IV show a statistically significant difference of .06 (s.e.=.032).

Finally, establishments where the union barely won also had a more compressed distribution of workers earnings prior to the election than establishments where the union barely lost. Figure VIII plots the 10th, 50th, and 90th percentiles of log pre-election earnings from the LEHD by non-overlapping 5-percent bins in the union vote share. For the lower-tail percentiles there is a substantial positive discontinuity at the threshold, while for upper-tail percentiles there is substantial negative discontinuity. Local linear estimates in Table IV show that the 10th percentile of pre-election earnings in establishments where the union barely won was .196 (s.e. = .024) log points higher than where the union barely lost, while the 90th percentile was .345 (s.e.=.054) log points *lower*. Establishments where the union barely won therefore had at baseline a much more compressed distribution of workers' earnings than where the union would eventually lose. This compression can be seen more succinctly in the difference between the 90th and 10th percentile of pre-election log earnings. Figure IX plots the 90-10 difference by union vote share, and shows a distinct and large drop at the threshold of union victory. The bottom row of results in Table IV shows the difference is about -.54 log points (s.e.=.034).

Interpreting these differences in light of the vote share discontinuities in the previous section, the evidence suggests that employers were more likely to be able to avert a close union victory (by vote challenges or otherwise) at smaller establishments and where workers earnings were more dispersed. Conversely, unions were more likely to be able to avert a close loss at larger establishments and where workers' earnings were more homogeneous.

5.2 Effects of union representation

The evidence of selection in close elections and the resulting discontinuities in pre-election characteristics of election establishments suggests that level comparisons of outcomes across the threshold of union victory that do not account for these differences would not consistently estimate the causal effect of union representation. Valid inference may still be possible, however, by exploiting the panel

nature of the data to account for pre-election differences between establishments on either side of the threshold of union victory. If the differences in potential outcomes between establishments on either side of the threshold are due to time-invariant factors, then using the local linear difference-in-differences specification (1) will consistently estimate the impact of unionization. These estimates suggest that unionization reduced the scale of production and shifted the composition of workers remaining at unionizing establishments.

5.2.1 Payroll, worker earnings, and employment

Evidence that takes into account pre-election selection suggests that union representation significantly reduced payroll, employment, and average worker earnings at unionizing establishments. Figure X plots the average year-to-year change in log payroll from the LBD by non-overlapping 5-percent bins in the union vote share. The left-hand panel corresponds to the year-to-year change in log payroll just prior to a union representation election, and the right-hand panel corresponds to the year-to-year change straddling the election. The left-hand panel shows no discernible jump in differenced pre-election log payroll at the 50 percent threshold. The continuity in the pre-election differences is consistent with minimal selection on log payroll differences, despite the presence of substantial selection on log payroll levels (cf. Figure VII), and lends credence to Assumption A1' as an alternative to the traditional RD identifying assumption A1. By contrast, the right-hand panel of Figure X (corresponding to pre- versus post-election differences) shows a substantial drop at the 50 percent threshold. Table V reports local linear difference-in-differences estimates corresponding to the figure for log payroll and other outcomes. The first row shows the estimated effect on log payroll is a highly significant $-.149$ (s.e. = $.018$), suggesting that a union victory had a substantial negative impact on establishment log payroll.

RD difference-in-differences estimates imply that the negative impact on payroll was driven by a decrease in average worker earnings, and to a smaller extent a decrease in employment. Figure XI plots differenced log average worker earnings by election vote share, with pre-election differences in the left panel and pre- versus post-election differences in the right panel. The right-hand panel shows a clear drop at the threshold. Table V quantifies the graphical evidence, suggesting that union representation led to a $.04$ (s.e.= $.005$) decrease in log average worker earnings. The pre-election difference shown in the left-hand panel of Figure XI is smaller, but appears to be somewhat positive,

suggesting that if anything the true magnitude of the effect on the post-election difference in worker earnings should be even larger. Employment, however, shows somewhat weaker evidence of a decline in response to a union victory. Figure XII plots averages of changes in log employment by union vote share for pre-election (left panel) and pre- versus post-election (right panel) differences. The left panel reassuringly shows no differences in employment differences at the 50 percent threshold. The right-hand panel shows a modest drop in employment at the threshold, suggesting that union representation reduced employment at establishments with close certification elections. Table V reports a marginally significant point estimate of $-.063$ (s.e.=.038) for the narrowest bandwidth, but the result is somewhat sensitive to bandwidth choice.

5.2.2 Worker composition

The effects on payroll, worker earnings, and employment are starkly counter to conventional wisdom on the effects of union representation. Theories of unionism suggest that unions may trade off pay increases with employment, but these results suggest otherwise. Does union representation really decrease worker pay without increasing employment? The answer is no. The reason is that these effects are driven by the union impact on the composition of workers. A union victory on average shifts workforce composition towards younger and lower-paid workers, with little impact on average earnings for workers who stayed.

Evidence from the LEHD suggests a union victory shifted the employee composition at an establishment toward younger workers. Figure XIII plots the average age of workers by union vote share both pre-election (left panel) and post-election (right panel). The left panel shows little evidence of selection on age, and the right panel shows a noticeable drop in average worker age at the 50 percent threshold. The top row of Table VI quantifies the drop at about -1.8 years (s.e.=.21).

The shift toward younger workers was driven by a relative increase in the hiring rate of younger workers in response to a union victory. Figure XIV plots the average age of employees who came to the establishment following a representation election by union vote share, and shows a substantial drop at the 50 percent threshold in the average age of workers leaving the establishment. Table VI shows the estimated effect is about -1.3 years (s.e.=.1).

The workers who were hired following a union victory were also on average lower paid. This can be seen by looking at the effect of union victory on the earnings of workers who came to the

establishment post-election. Figure XV plots median log annual earnings of workers who came following a union election by the union vote share. The plot shows a substantial drop at the 50 percent threshold, implying that a union victory increased the relative hiring rate of lower-paid workers. The third row of results in Table VI estimates a highly significant difference of $-.211$ log points (s.e. = $.052$).

The shift toward lower-paid workers was also driven by increased exit of higher-paid workers. This can be seen by looking at the effect of union victory on the pre-election earnings of workers who leave the establishment after a union election. Figure XVI plots average log annual earnings of workers who leave following a union election by the union vote share. The plot shows a substantial jump at the 50 percent threshold, implying that a union victory increased the relative exit rate of higher-paid workers. The fourth row of results in Table VI estimates a large and highly significant difference of $.585$ (s.e. = $.088$).

Much of the negative effect on payroll and average worker earnings, then, is largely driven by the effect of unionization on who leaves and who comes in response to a union victory. Younger, lower-paid workers are more likely to be hired following a union victory, while higher-paid workers are more likely to leave. The result is that average worker earnings at establishments where the union won an election are lower.

But were the earnings of workers who stayed affected by union representation? Figure XVII plots the average pre- versus post-election change in log earnings for workers who remained employed at election establishments by the union vote share. The figure shows little evidence of a substantial effect on the earnings of those who stayed. Table VI confirms the graphical evidence, with the corresponding estimate of $.003$ (s.e.= $.016$). On average, there is little evidence that union representation had a large impact on the earnings of workers who remained at election establishments.

5.2.3 Interpretation

The single largest establishment-level impact of union representation appears to be the effect on what kinds of workers end up employed at unionized plants with little impact on the average earnings of workers who stay. Together with findings elsewhere that unions compress the distribution of individual-level earnings (Frandsen, 2012), the evidence is consistent with models of unionization

that imply that the set of workers at unionized plants is determined by two-sided selection on the part of both the employer and potential workers (Abowd and Farber, 1982; Card, 1996).

A simplified example of this kind of model, adapted from Card (1996), endows a worker i with general productivity g_i , which is also her wage in the nonunion sector:

$$w_i^n = g_i.$$

The return to skill in the union sector, however, is lower, so the worker's wage in a union job would be:

$$w_i^u = \theta_0 + \theta_1 g_i,$$

where $\theta_0 > 0$ and $0 < \theta_1 < 1$. After a successful unionizing drive, workers decide to stay at the union plant (if they were there already) or queue for a union job (if they weren't) if the difference between the union and nonunion wage exceeds ρ_i , the person-specific preference cost of working in a union job. The worker's decision to stay or join is therefore determined by:

$$g_i < \frac{\theta_0}{1 - \theta_1} - \frac{\rho_i}{1 - \theta_1}. \quad (2)$$

The newly-unionized employer decides to keep a worker (or hire her from the queue) if the sum of the worker's general productivity and match-specific productivity ω_i exceeds the union wage. The employer's criterion to keep or hire a worker is therefore:

$$g_i > \frac{\theta_0}{1 - \theta_1} - \frac{\omega_i}{1 - \theta_1}. \quad (3)$$

The worker's and the employer's decision rules have opposing influences on the composition of workers who end up employed at the establishment: given the depressed return to skill, while the employer would prefer to hire and retain higher-skilled workers, they prefer to leave in response to a union victory. The model predicts that when the shift in the union wage scale (captured by θ_0) is small, the worker's decision rule (2) is more likely to bind, shifting the composition of workers who end up employed at the unionized plant toward the lower end of the productivity distribution.

The evidence in this paper supports this prediction. The lack of an overall effect on the earnings

of workers who remained at the establishment post-election suggests θ_0 is quite small. Consistent with the model's prediction, both the finding that workers who join following a union victory are younger and lower paid and the finding that workers who leave establishments following a union victory are higher-paid suggest that at the hiring and exit decisions the employee selection rule is more binding. The model also suggests an interpretation of the employment effects reported above. In terms of the model, when θ_0 is small, the fraction of workers who prefer to stay following a union victory will also be small. The modest negative effect on employment may therefore reflect primarily voluntary separations rather than firings in response to the union victory.

6 Summary and Conclusions

This paper presented evidence of nonrandom selection in close union elections. The distribution of election outcomes showed surprisingly few union victories when discarding a single vote would have reversed the outcome in favor of the employer, and likewise when the outcome could have been reversed in favor of the union. The potential manipulation of close elections led to significant discontinuities in the underlying characteristics of establishments at the majority-rule threshold of union victory, casting doubt on causal estimates of union impacts based on level comparisons of establishments where the union barely won or barely lost.

Estimates exploiting the panel nature of the data to account for pre-election selection are quite different from previous results. They suggest that union representation led to a decrease in establishment payroll, employment, and average worker earnings. The decrease in payroll and earnings was primarily driven by changes in workforce composition in response to a union victory. Higher-paid workers were more likely to leave after a union victory, and younger, lower-paid workers were relatively more likely to come. Thus the remaining pool of workers was younger and lower paid. There is no evidence of any impact on the earnings of workers who remained at the plant. Thus, the single largest effect of union representation in these results is the impact on the mix of workers remaining employed at the establishment, consistent with models of unionization in which the set of workers at union plants is subject to selection on the part of both the employer and employees.

As is generally true of discontinuity-based research designs, the causal impacts reported here

apply to establishments that were near the margin of union victory or defeat. The modal representation election is close, so in this sense the findings here apply to typical unionizing drives, but certainly not to all. Apart from differences in sample and research design, this could partially explain the contrast between the sizeable union premium found in much of the previous literature and the lack of evidence of an individual earnings effect in this study.

These findings also differ from previous election-based RD studies of union effects, which found little impact on a range of establishment-level outcomes. Part of the difference is explained by pre-election selection in close union elections that the evidence here accounts for. The evidence here also includes a broader set of industries, a wider time frame, and better name-and-address match quality, which may also explain the contrasting results.

The findings reported here sound a cautionary note, but perhaps not a death knell for RD designs based on close elections, even when there is evidence for selection. First, the findings here underscore the need for pre-treatment “balance” tests and specification checks. But the very panel data used to detect potentially confounding selection can also be used to construct estimators that account for pre-existing differences in a hybrid of regression discontinuity and difference-in-difference research designs.

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Appendix

Construction of the dataset

As described in the text, the dataset used in this paper consists of NLRB certification election results matched to confidential establishment-level datasets maintained by the Census Bureau.

The union certification election records were collected by the NLRB, and in large part maintained by the AFL-CIO. John-Paul Ferguson, Thomas Holmes, and Hank Farber obtained the election records from the NLRB, and made them available for this research. The complete data

set covers the period 1963-2009, although the main sample used in the analysis covers the years 1980-2009. The raw data contains results from elections stemming from several different type of petitions, including cases where a union seeks to be certified (RC), an employer seeks an election to remove an existing union (RM), or employees seek to remove a union (RD). I restrict to RC-cases, where a union seeks certification. The dataset contains many duplicate records. In some cases they are true duplicates: one election generated multiple records in the database. In these cases I simply delete the redundant entries. In other cases, multiple entries arise from more than one union being on the ballot. In these cases the relevant union vote share is the largest one; I therefore retain the entry with the largest vote share, and delete the others. Finally, in some cases multiple elections were held at the same establishment because, for example, different groups of workers constituted different bargaining units. Since I can't distinguish between workers in different bargaining units, the relevant vote share is the largest, so again I keep only the entry corresponding to the election where the union received the highest vote share.

The second data component consists of the Census Bureau's Longitudinal Business Database (LBD). As described in the text, the LBD contains longitudinally-linked establishment-level panel data on payroll and employment for virtually the universe of U.S. private-sector employers. The names and addresses for these establishments are available from the Standard Statistical Establishment List (SSEL), also known as the Business Register (BR), which provides the sampling frame for various economic censuses and surveys.

The matching process to combine these two data sources is as follows. First, employer name and address information from both the NLRB dataset and the Census Bureau's Business Register (BR) were cleaned and standardized using the SAS Data Quality Server standardization functions. NLRB election records were then matched to BR records by several combinations of state, county, city, employer name, street address, and industry code. The match was performed iteratively in descending order of strictness. The cutoff level of strictness was determined by hand checking matches from each iteration, and stopping once match quality dipped below 95 percent. This procedure successfully matched 82 percent of the NLRB election records, much higher than the 26 percent match rate in DiNardo and Lee (2004).

Finally, some of the analysis required individual-level data on employees at election establishments. Individual-level earnings data came from the Employment History Files (EHF) within the

LEHD database. The EHF contains employee, employer, and earnings data for each employment relationship that generated at least one dollar of wages. The EHF includes a state employer identification number (SEIN) with each record, and in some cases an identifier for the establishment within the employer, which is important for multi-unit employers. For the cases where there is no establishment identifier, the LEHD provides a Unit-to-Worker (U2W) imputation to assign workers to establishments. The Census Bureau maintains a crosswalk of firm- and establishment-level identifiers (the Federal Employer Identification Number, or EIN, county, and industry) which was used to merge the election establishments identified in the LBD with the LEHD. The overall match rate between the NLRB records that correspond to states and years where the LEHD is available was only slightly lower (77 percent).

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Table I: NLRB Certification Elections Summary Statistics

	All	Union loss	Union victory
Pr(Union victory)	45%	0%	100%
Union vote share, average	49%	33%	70%
Number of voters, average	93	100	84
Number of voters, total	4,191,075	2,494,768	1,696,307
Number of elections	45,176	24,974	20,202

Data source: NLRB certification election records.

Table II: Average Establishment Characteristics from the Longitudinal Business Database

	All	Union Loss	Union Victory
Number of employees			
<i>pre-election</i>	254	236	277
<i>post-election</i>	259	242	281
Payroll (millions)			
<i>pre-election</i>	\$8.87	\$8.27	\$9.59
<i>post-election</i>	\$9.13	\$8.62	\$9.75
Average worker earnings			
<i>pre-election</i>	\$36,100	\$35,511	\$36,803
<i>post-election</i>	\$35,219	\$36,319	\$33,903

Notes: Payroll and worker earnings in year 2000 dollars.

Table III: Average Employee Characteristics from the LEHD

	All	Union Loss	Union Victory
Age at time of election	40	39	41
female	44%	42%	45%
Nonwhite	38%	35%	40%
Quarterly earnings	\$8,097	\$8,204	\$7,975
Stay at establishment following election	42%	43%	41%
N	1,774,975	946,607	828,368

Notes: Pre-election means. Quarterly earnings are in year 2000 dollars.

Table IV: Discontinuities in pre-election establishment and worker characteristics

Dependent Variable:	(1)	(2)	(3)
log employment	0.134 (0.032)	0.129 (0.037)	0.228 (0.062)
log payroll	0.060 (0.032)	0.059 (0.043)	0.133 (0.045)
log worker quarterly earnings percentile:			
10th	0.196 (0.024)	0.240 (0.034)	0.350 (0.111)
20th	0.156 (0.023)	0.190 (0.030)	0.282 (0.091)
50th	0.047 (0.031)	0.102 (0.040)	0.133 (0.085)
80th	-0.081 (0.037)	-0.019 (0.039)	-0.019 (0.074)
90th	-0.345 (0.054)	-0.308 (0.032)	-0.368 (0.057)
90-10 difference	-0.541 (0.034)	-0.548 (0.052)	-0.718 (0.136)
bandwidth	0.15	0.2	0.25
local polynomial type	linear	linear	quadratic

Notes: Estimates and robust standard errors for the coefficient on an indicator for union victory in local polynomial regressions of the dependent variable in the left-hand column on an indicator for union victory interacted with a polynomial of the indicated degree in the union vote share. Local regressions restrict to observations within the indicated bandwidth on either side of the 50 percent threshold. The data were collapsed by the union vote share variable.

Table V: Effects of union representation on establishment outcomes

Dependent Variable:	(1)	(2)	(3)
ln(payroll)	-0.149 (0.018)	-0.177 (0.028)	-0.148 (0.028)
ln(employment)	-0.063 (0.038)	-0.125 (0.030)	-0.066 (0.056)
ln(avg worker earnings)	-0.040 (0.005)	-0.038 (0.003)	-0.043 (0.006)
bandwidth	0.15	0.2	0.25
local polynomial type	linear	linear	quadratic

Notes: Estimates and robust standard errors for the coefficient on an indicator for union victory from local polynomial regressions of the differenced dependent variable indicated on an indicator for union victory interacted with a polynomial of the indicated degree in the union vote share. The differenced dependent variable is the observation from the year following the election minus the observation from the year prior to the election. The data were collapsed by the union vote share variable.

Table VI: Effects on Worker Composition and Stayer's earnings, LEHD

Dependent Variable:	(1)	(2)	(3)
age (all)	-1.842 (0.214)	-1.507 (0.187)	-1.687 (0.300)
age (comers)	-1.335 (0.104)	-1.559 (0.140)	-1.651 (0.280)
log quarterly earnings (comers)	-0.211 (0.052)	-0.193 (0.029)	-0.235 (0.058)
log quarterly earnings (leavers)	0.585 (0.088)	0.478 (0.107)	0.764 (0.061)
log quarterly earnings (stayers, differenced)	0.003 (0.016)	0.002 (0.019)	0.004 (0.019)
bandwidth	0.15	0.2	0.25
local polynomial type	linear	linear	quadratic

Notes: Estimates and robust standard errors for the coefficient on an indicator for union victory in local polynomial regressions of the dependent variable in the left-hand column on an indicator for union victory interacted with a polynomial of the indicated degree in the union vote share. The differenced log quarterly earnings of stayers consists of the individual's average quarterly earnings in the year following the election minus the individual's average quarterly earnings in the year prior to the election. Local regressions restrict to observations within the indicated bandwidth on either side of the 50 percent threshold. The data were collapsed by the union vote share variable.

Table VII: Match success rates between NLRB and Census datasets

	<u>Number of elections</u>	<u>Percent</u>
NLRB dataset, all	88,356	100%
Matched to LBD	72,424	82%
NLRB dataset, LEHD states and years	26,307	100%
Matched to LEHD	20,160	77%

Notes: Match frequencies and rates from name-and-address matching procedure between the NLRB elections dataset and the Census datasets (LBD and LEHD). The match rate denominator for matching to the LEHD includes only elections in 30 states covered by the LEHD during the years in which data are available.

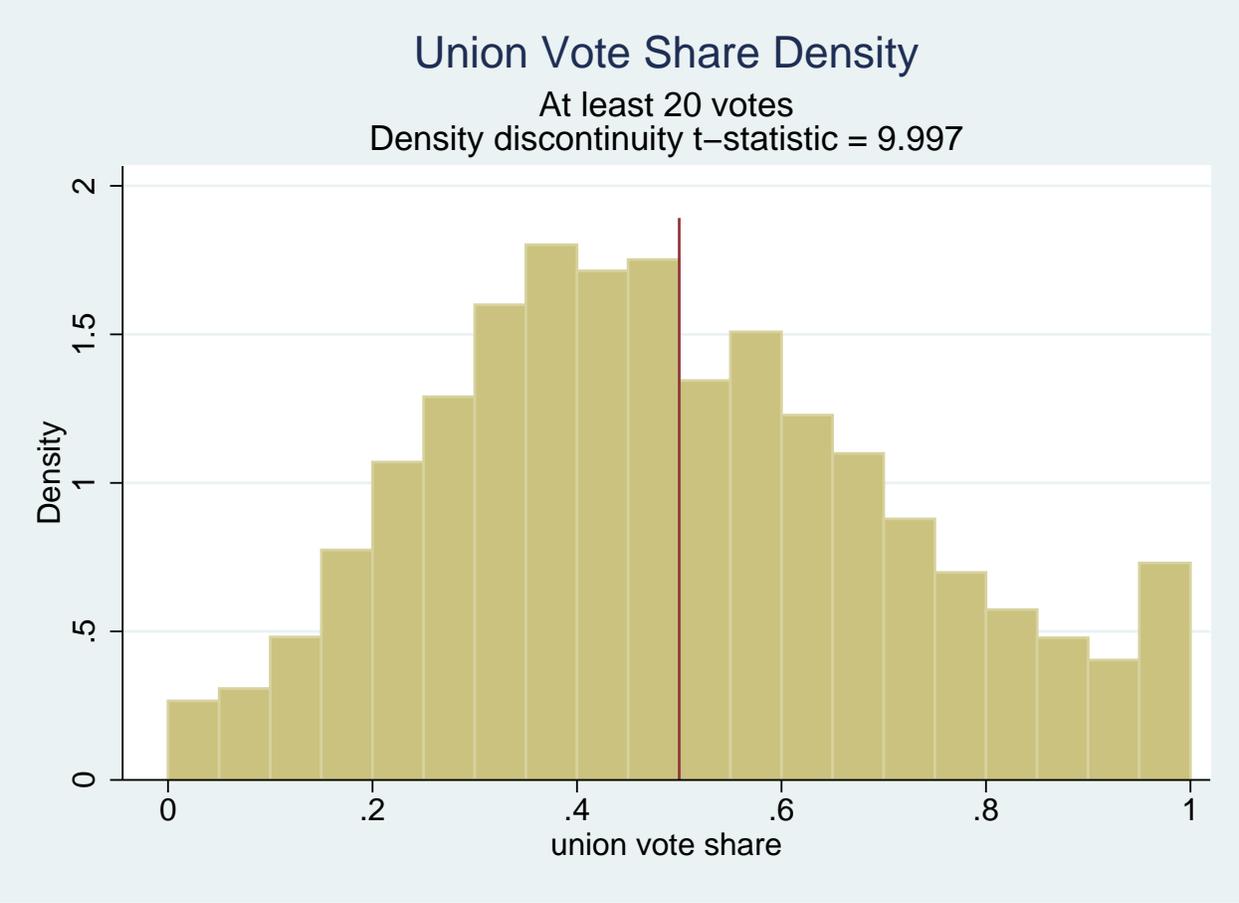


Figure I: Density of binned union vote share for elections with at least 20 voters. The density discontinuity t-statistic is from a McCrary (2008) test using a triangular kernel with a bandwidth of .3. Data are from NLRB election records.

Union Vote Share Density by Election Size

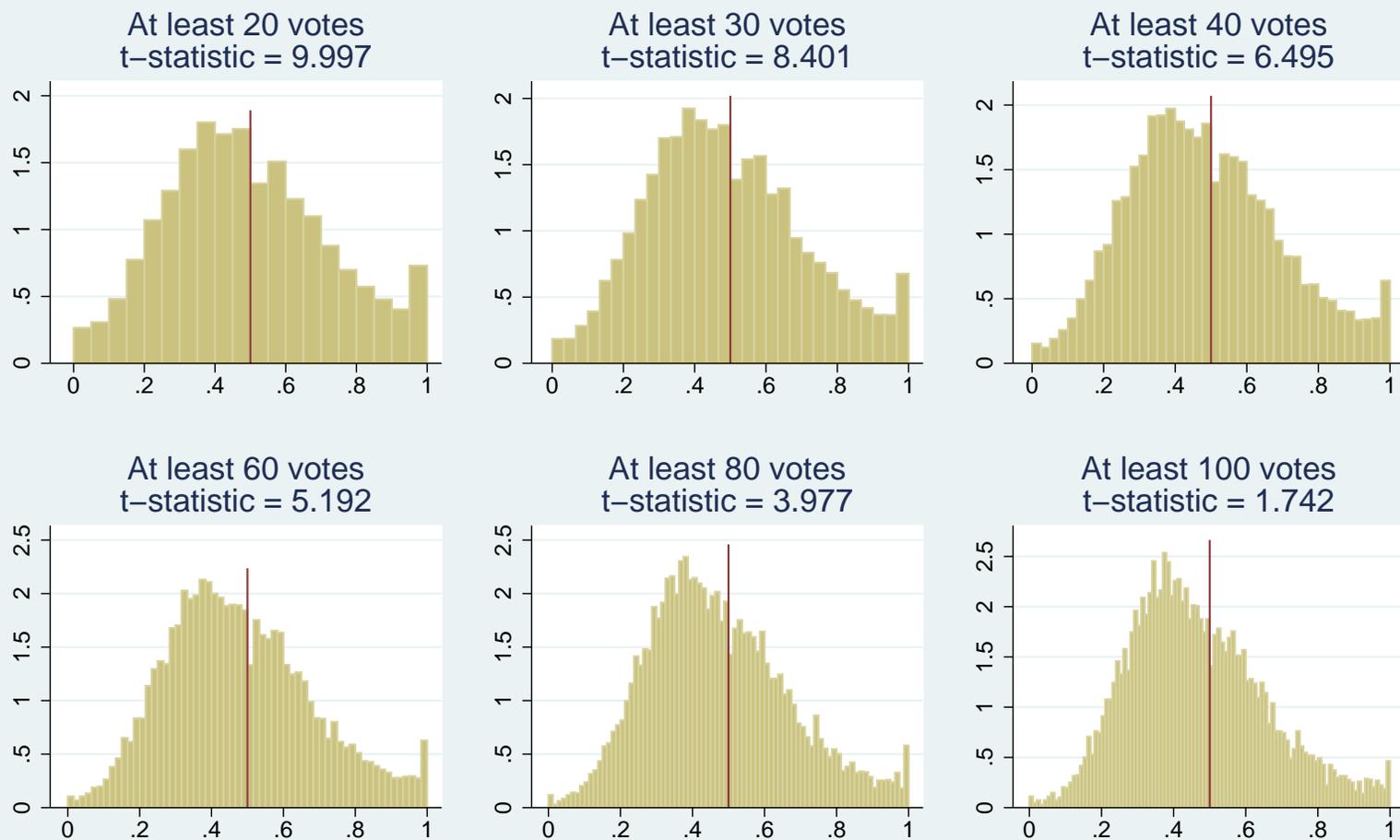


Figure II: Density of binned union vote share for elections with at least the indicated number of voters. The reported t-statistics are from McCrary (2008) density discontinuity tests using a triangular kernel with a bandwidths of .1 to .3. Data are from NLRB election records.

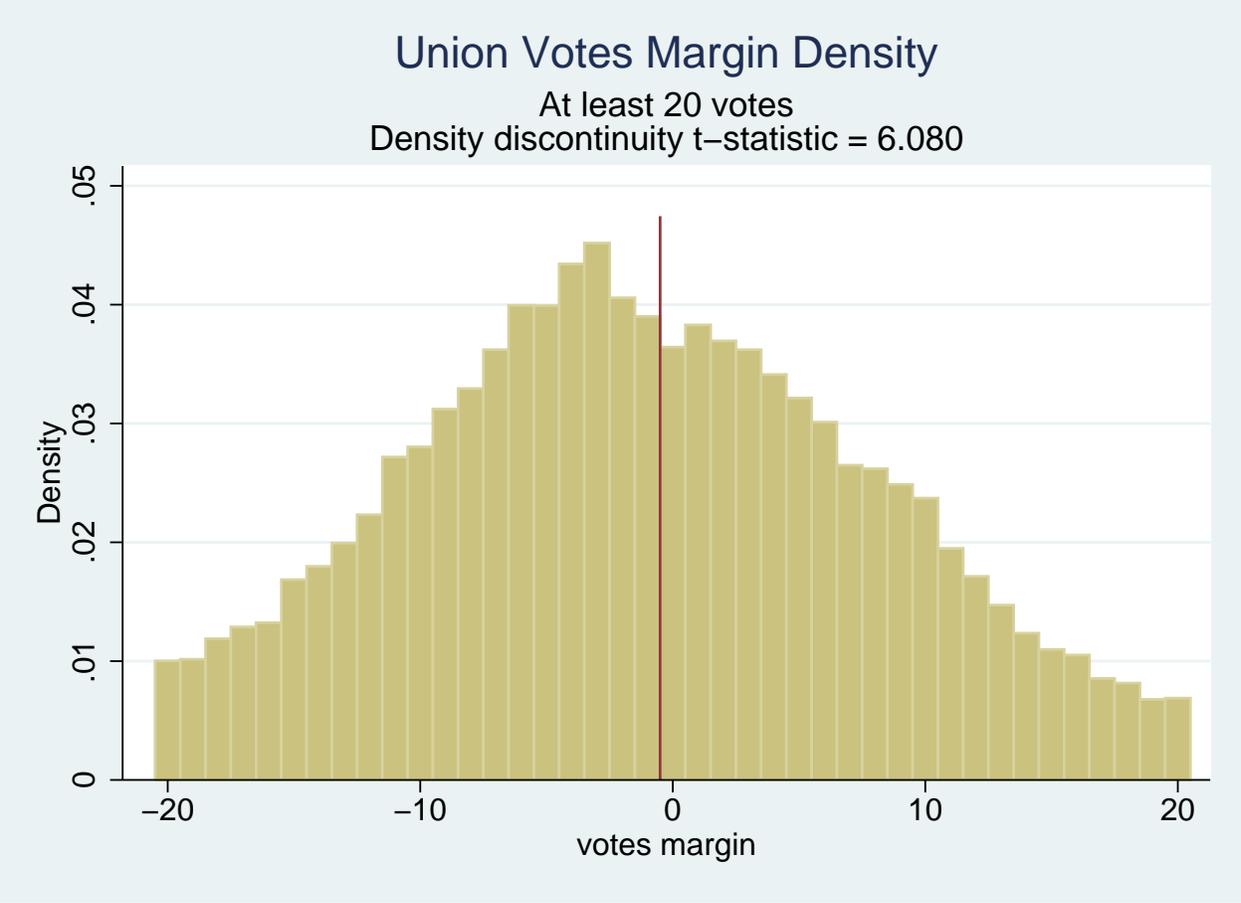


Figure III: Density of the union votes margin for elections with at least 20 voters. The density discontinuity t-statistic is from a McCrary (2008) test using a triangular kernel with a bandwidth of 15 votes. Data are from NLRB election records.

Union Votes Margin Density, Odd number of voters

At least 20 votes

Density discontinuity t-statistic = 6.363

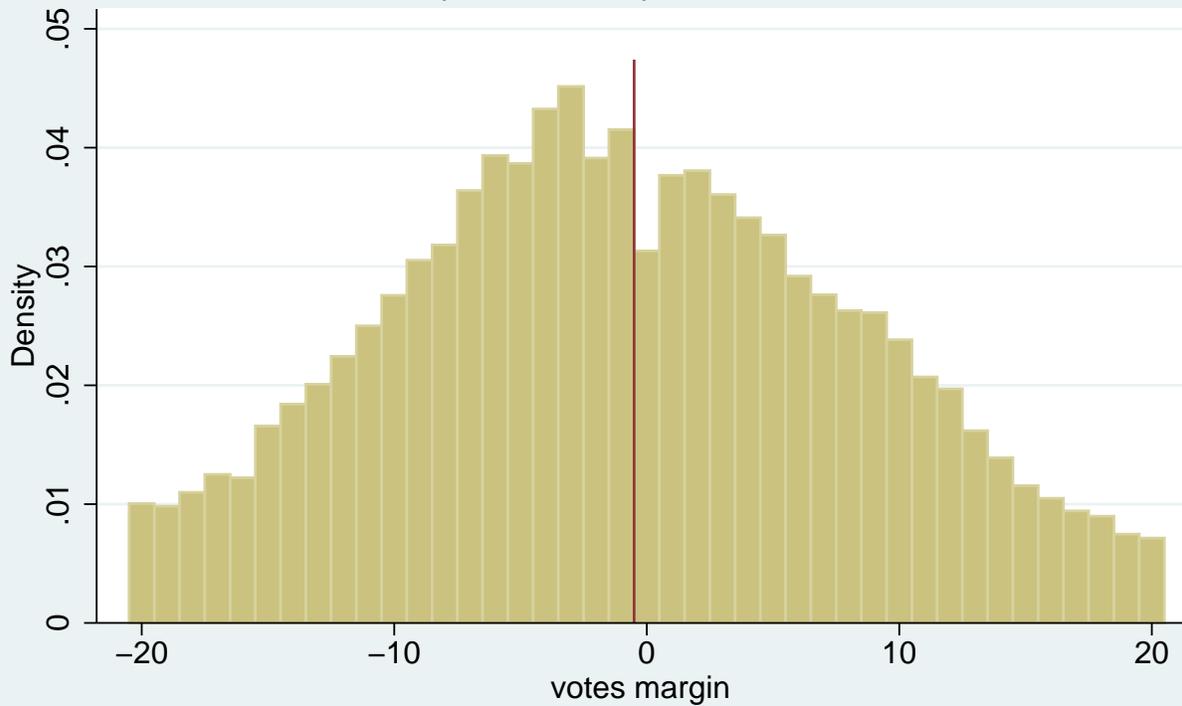


Figure IV: Density of the union votes margin for elections with an odd number of voters and at least 20 votes. The density discontinuity t-statistic is from a McCrary (2008) test using a triangular kernel with a bandwidth of 15 votes. Data are from NLRB election records.

Union Votes Margin Density, Even number of voters

At least 20 votes

Density discontinuity t-statistic = 2.539

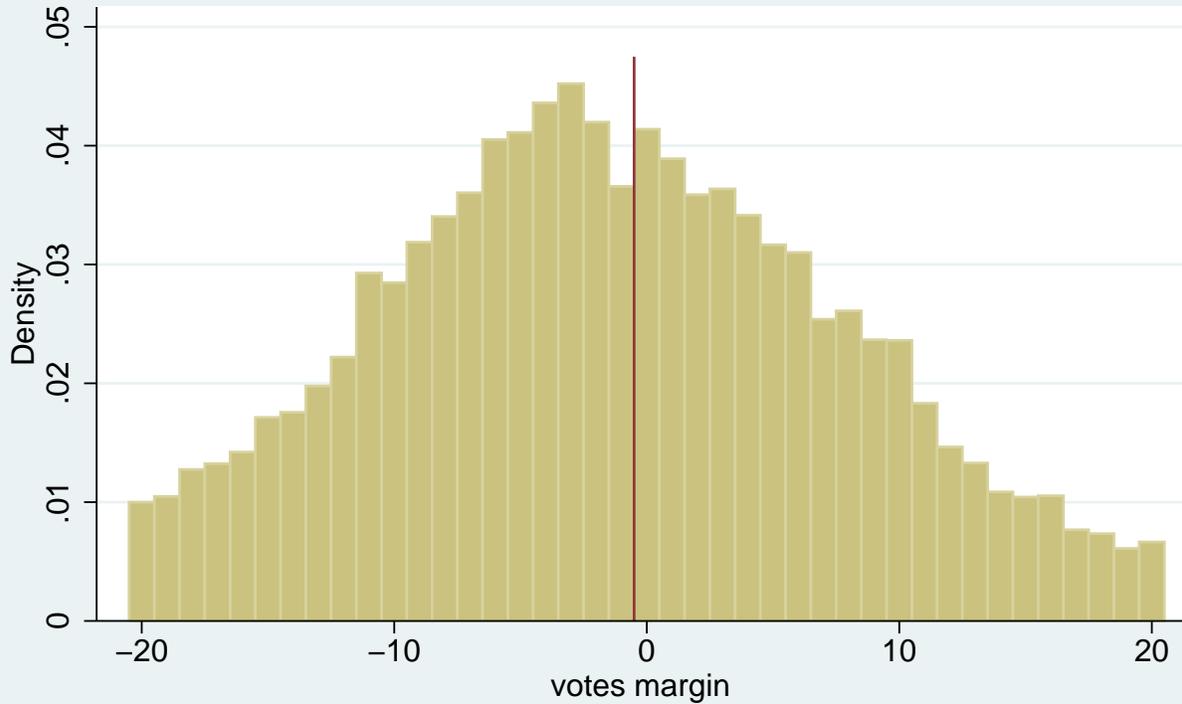


Figure V: Density of the union votes margin for elections with an even number of voters and at least 20 voters. The density discontinuity t-statistic is from a McCrary (2008) test using a triangular kernel with a bandwidth of 15 votes. Data are from NLRB election records.

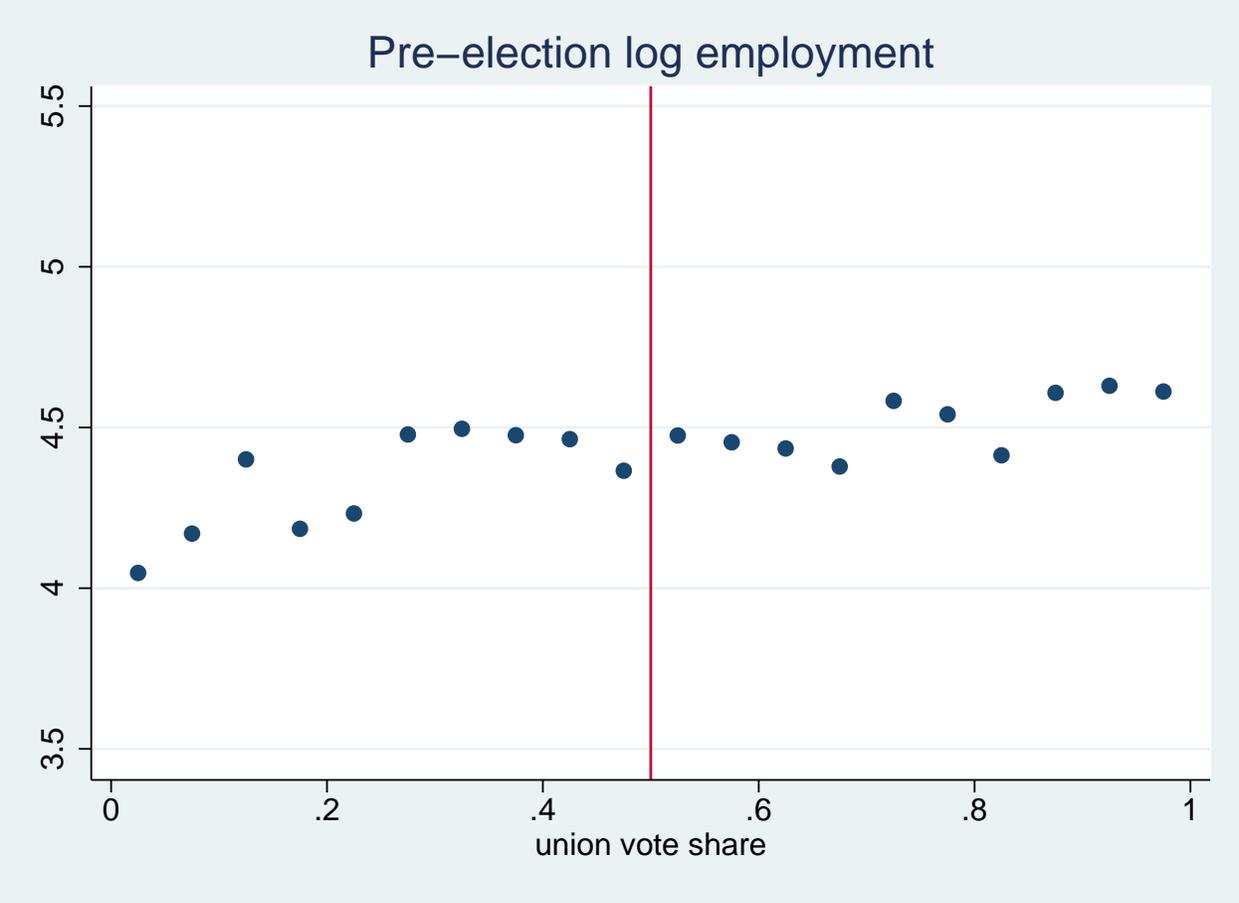


Figure VI: Pre-election log employment by union vote share. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LBD and NLRB election records.

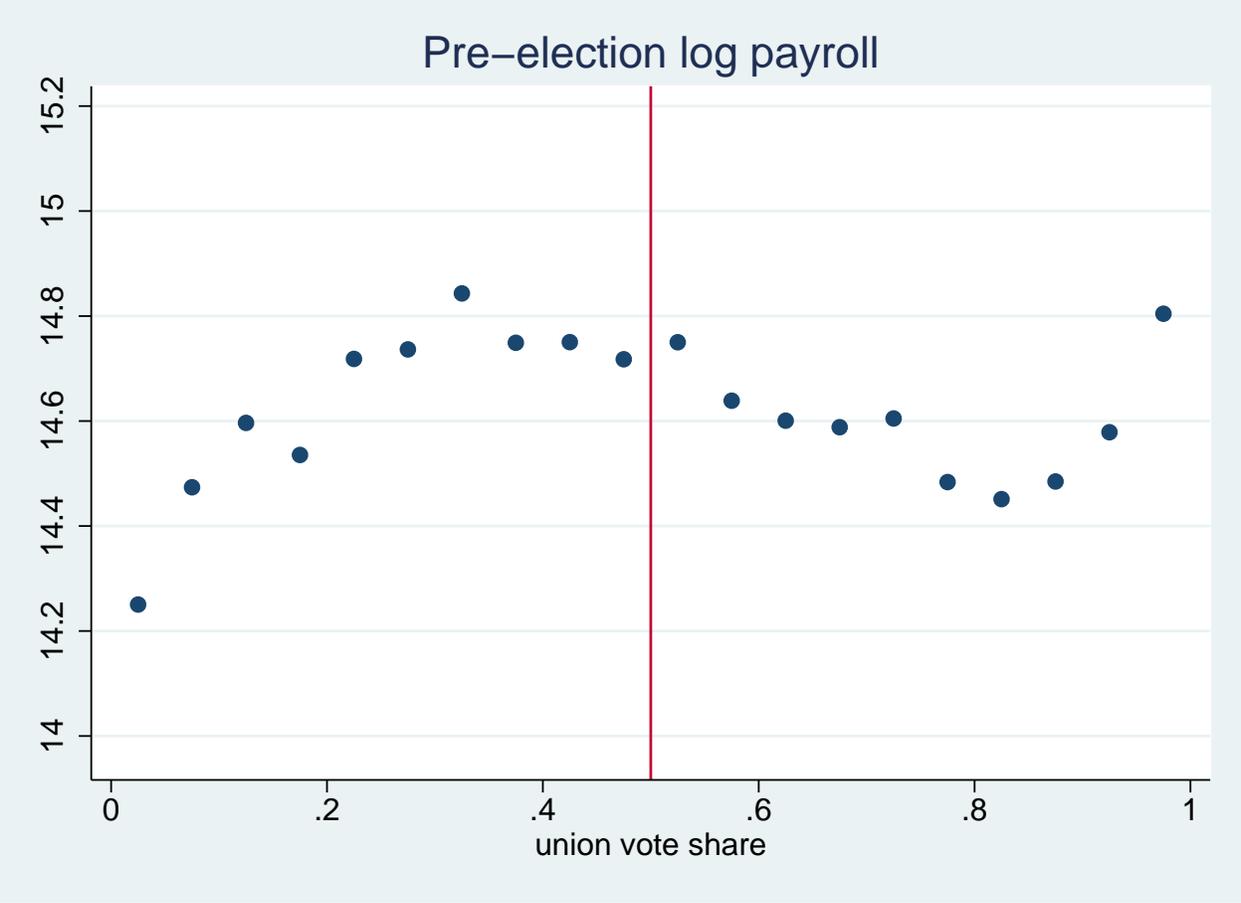


Figure VII: Pre-election log payroll by union vote share. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LBD and NLRB election records.

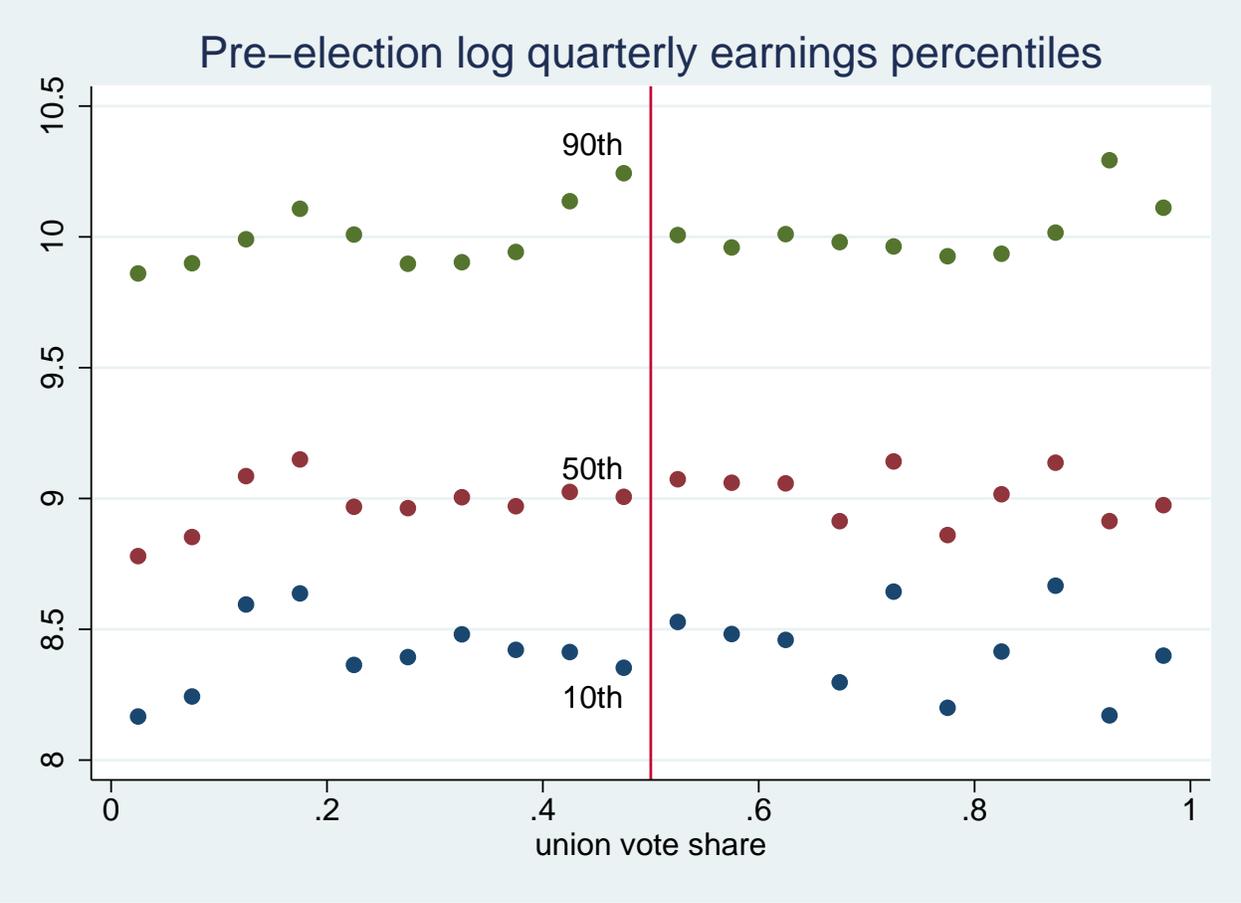


Figure VIII: 90th, 50th, and 10th percentiles of pre-election log quarterly earnings by union vote share RD graph. The points are raw percentiles by non-overlapping vote share bins of width .05. Data are from the LEHD and NLRB election records.

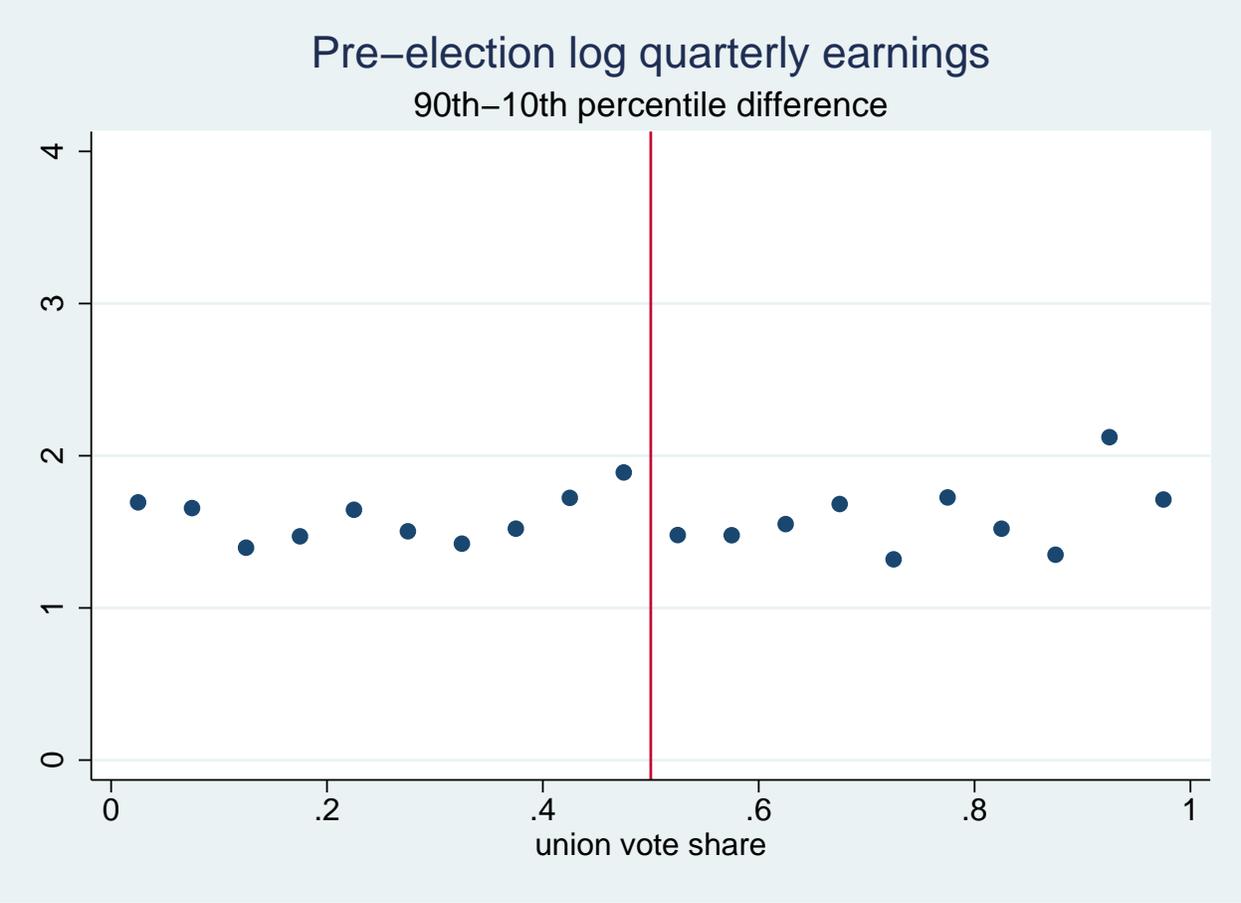


Figure IX: Difference between the 90th and 10th percentile of pre-election log quarterly earnings by union vote share. The points are differences in raw percentiles by non-overlapping vote share bins of width .05. Data are from the LEHD and NLRB election records.

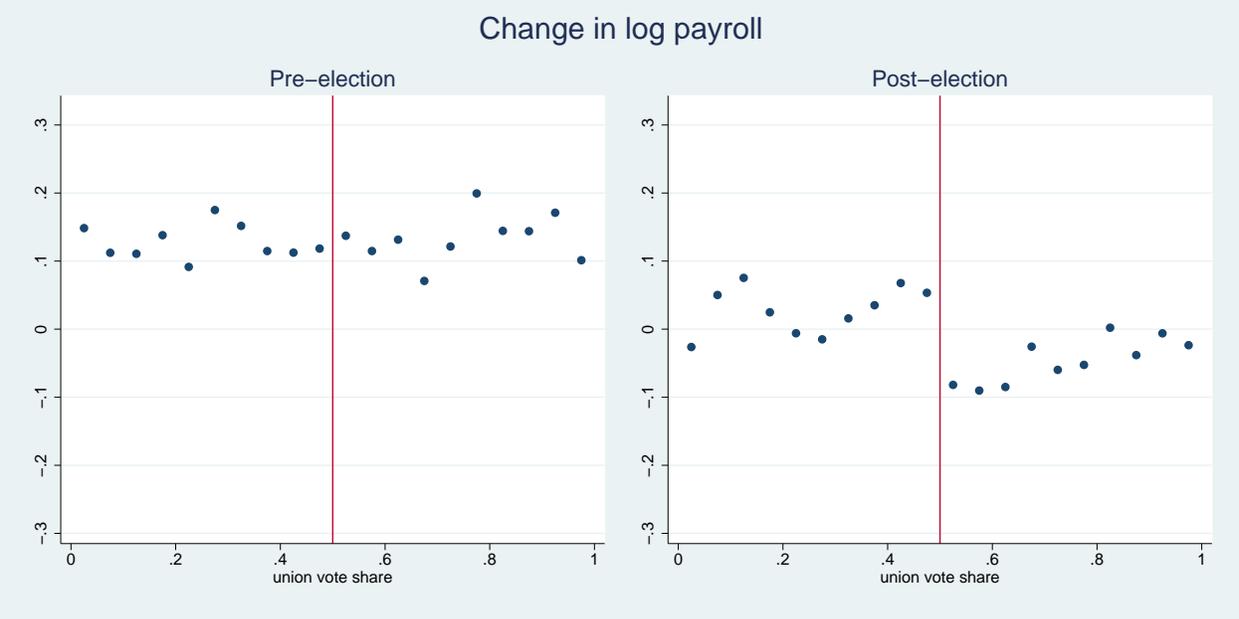


Figure X: Change in log payroll by union vote share. The left panel is the change from two years prior to the election to one year prior to the election. The right panel is the change from the year prior to the election to the year after the election. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LBD and NLRB election records.

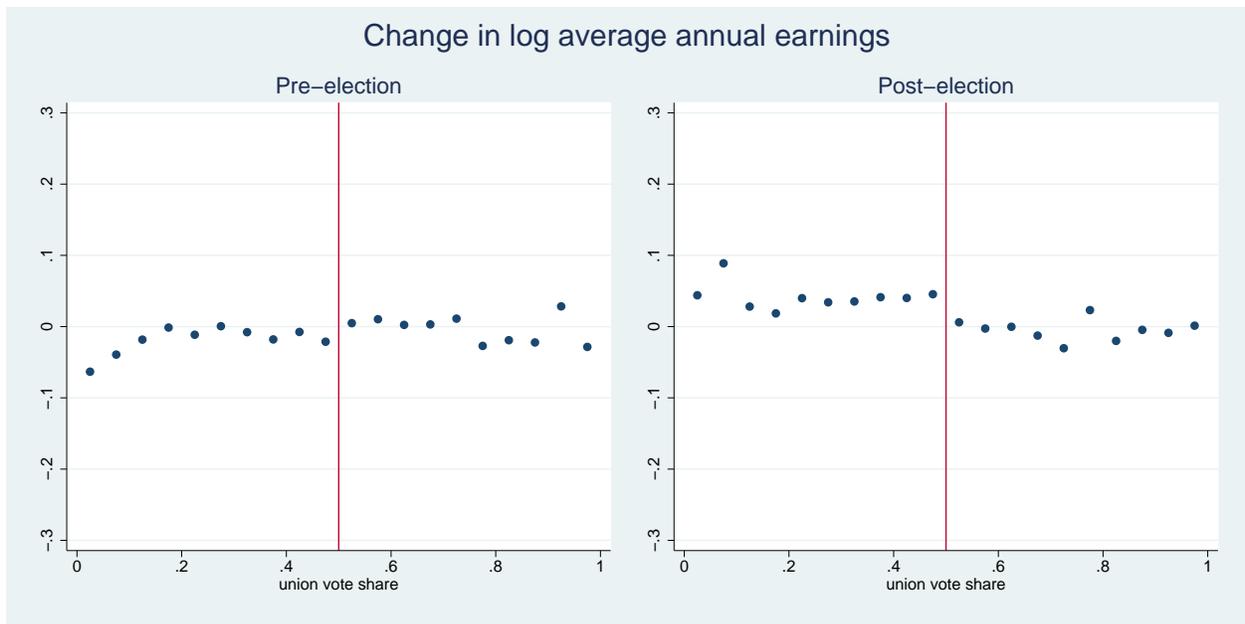


Figure XI: Change in log average annual earnings by union vote share. The left panel is the change from two years prior to the election to one year prior to the election. The right panel is the change from the year prior to the election to the year after the election. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LBD and NLRB election records.

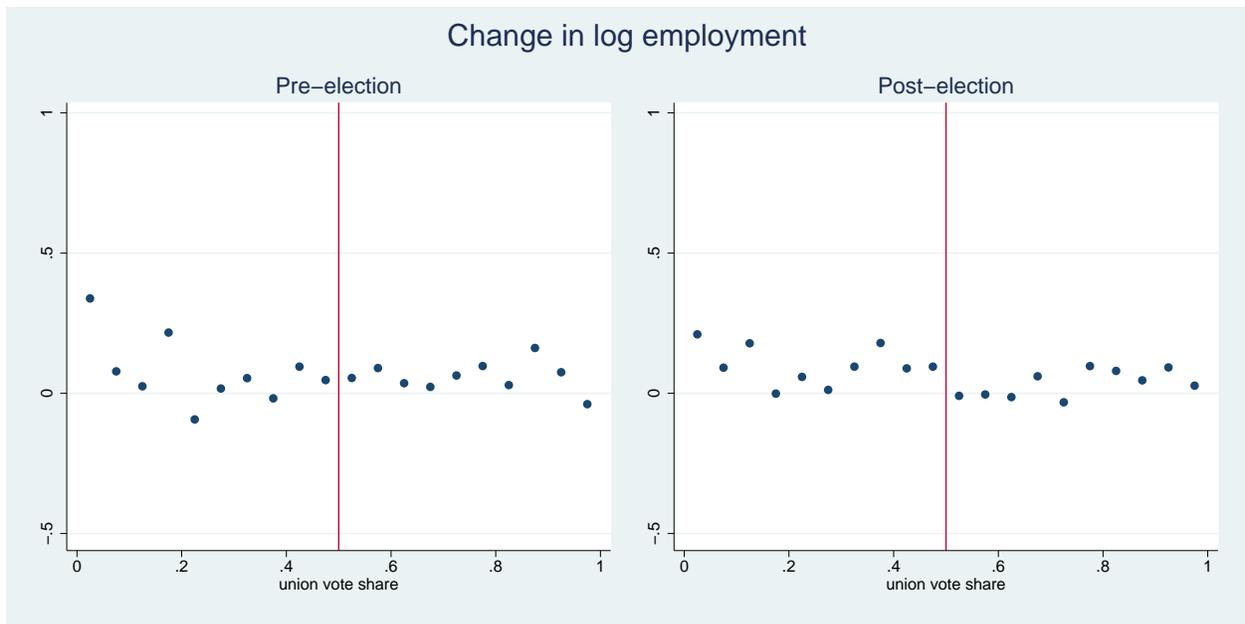


Figure XII: Change in log employment by union vote share. The left panel is the change from two years prior to the election to one year prior to the election. The right panel is the change from the year prior to the election to the year after the election. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LBD and NLRB election records.

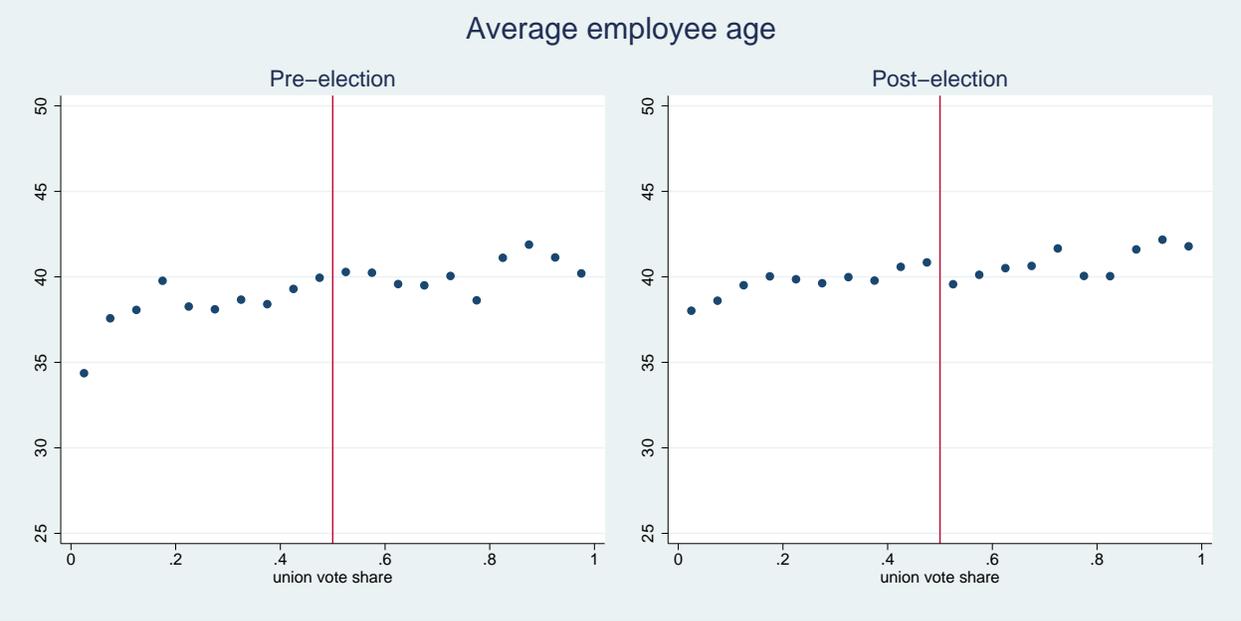


Figure XIII: Average employee age by union vote share. The left panel includes employees during the year prior to the election. The right panel includes employees during the year following the election. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LEHD and NLRB election records.

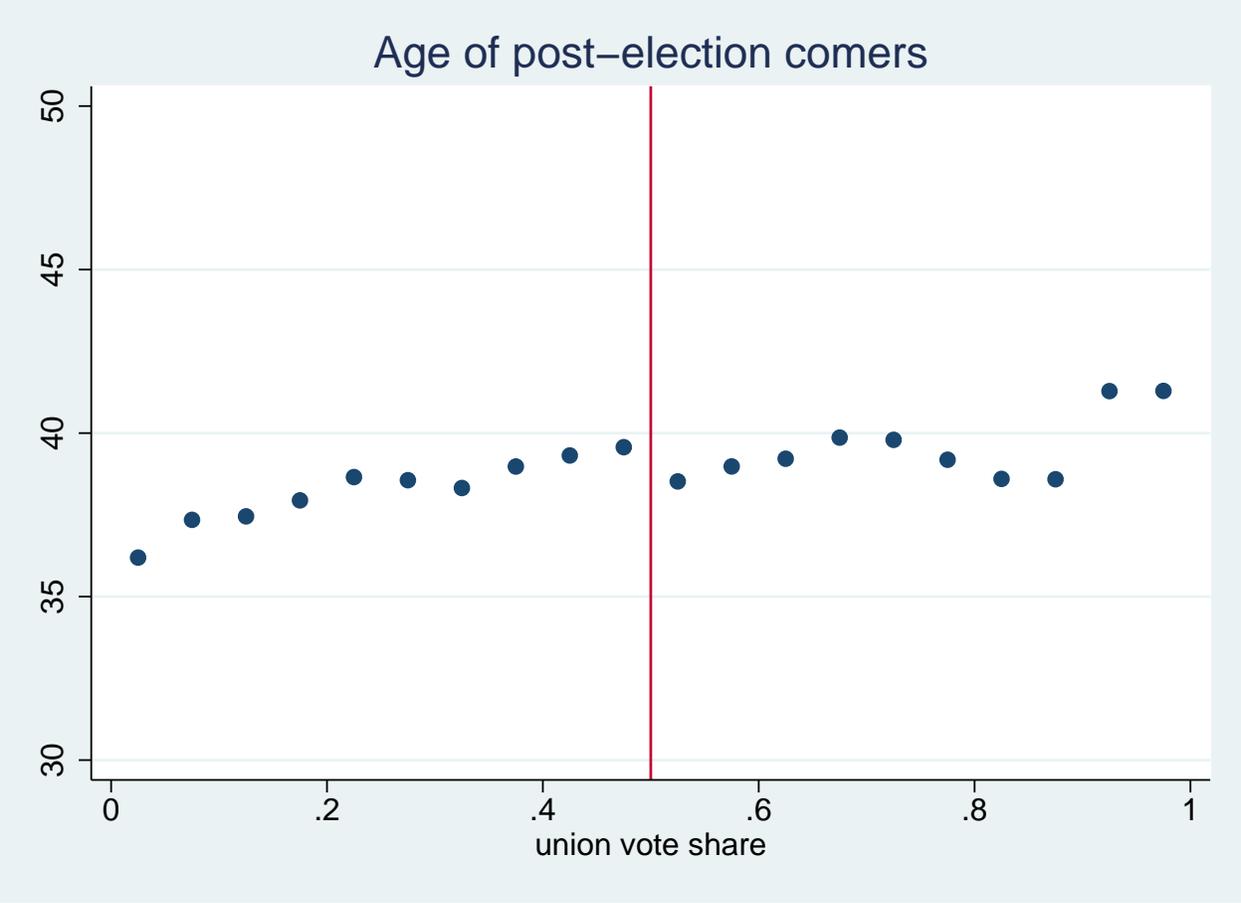


Figure XIV: Average age of employees who came to an establishment post-election by union vote share. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LEHD and NLRB election records.

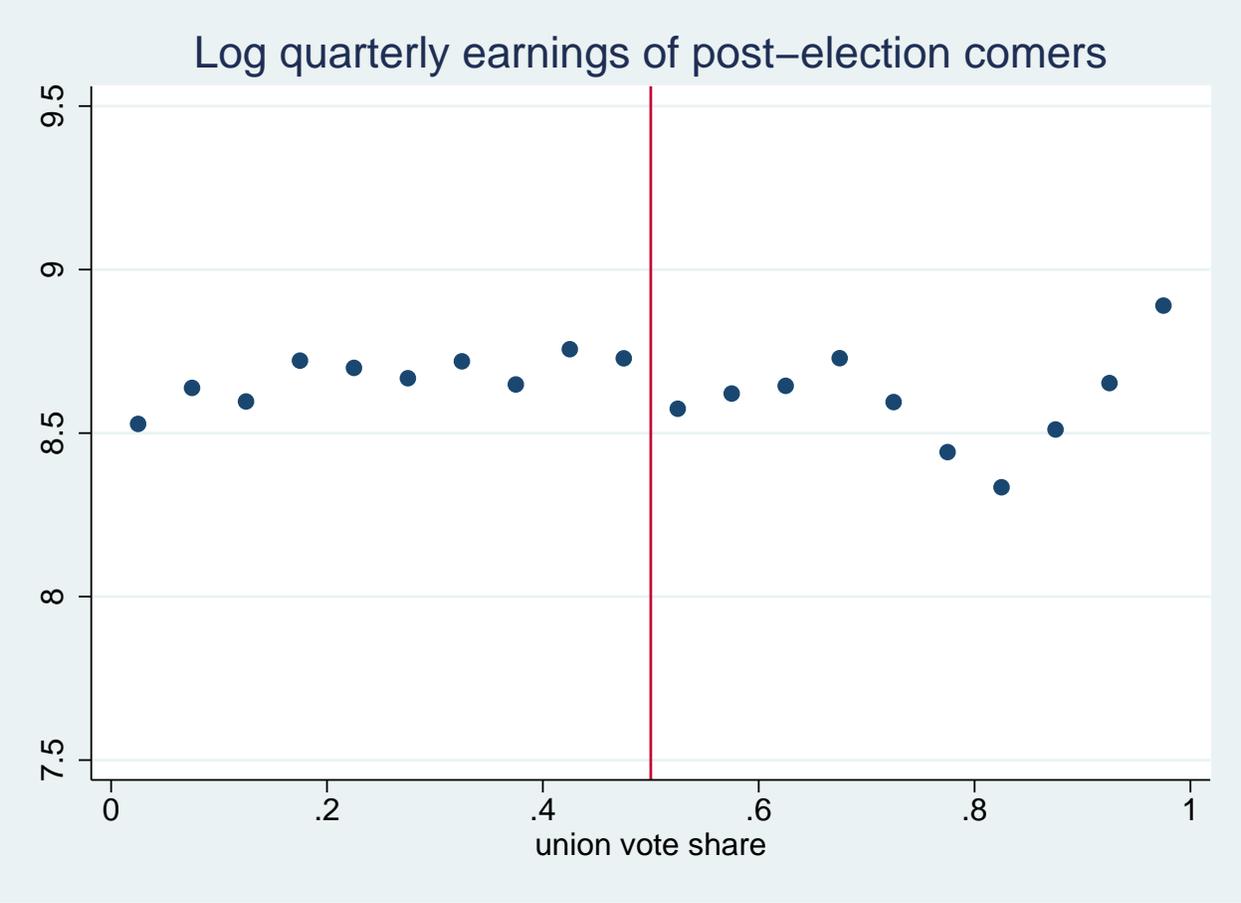


Figure XV: Log quarterly earnings of workers who came to an establishment post-election. The points are raw medians by non-overlapping vote share bins of width .05. Data are from the LEHD and NLRB election records.

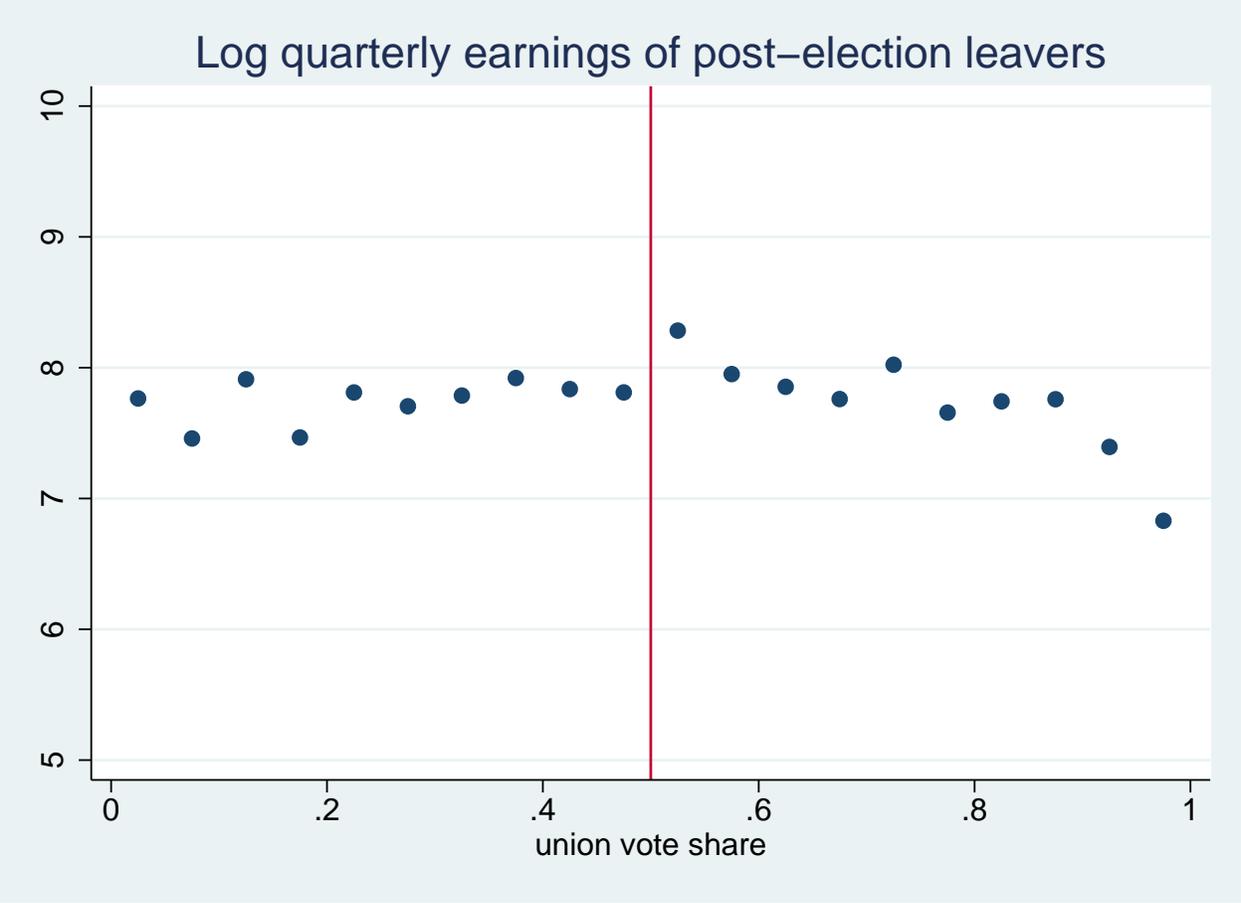


Figure XVI: Average log pre-election quarterly earnings of leaving employees by union vote share. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LEHD and NLRB election records.



Figure XVII: Change in log quarterly earnings of employees who stay by union vote share. The left panel is the change from two years prior to the election to one year prior to the election. The right panel is the change from the year prior to the election to the year after the election. The points are raw averages by non-overlapping vote share bins of width .05. Data are from the LBD and NLRB election records.