The Impact of Unions on Nonunion Wage Setting: Threats and Bargaining

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Abstract

In this paper we provide new estimates of the impact of unions on nonunion wage setting. We allow the presence of unions to affect nonunion wages both through the typically discussed channel of nonunion firms emulating union wages in order to fend off the threat of unionisation and through a bargaining channel in which nonunion workers use the presence of union jobs as part of their outside option. We specify these channels in a search and bargaining model that includes union formation and, in our most complete model, the possibility of nonunion firm responses to the threat of unionisation. Our results indicate an important role played by union wage spillovers in lowering wages over the 1980-2010 period. We find de-unionisation can account for 38% of the decline in the mean hourly wage between 1980 and 2010, with two-thirds of that effect being due to spillovers. Both the traditional threat and bargaining channels are operational, with the bargaining channel being more important.

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1 Introduction

Private sector unionisation in the United States is very nearly dead. In 2019, only 6.2% of private sector workers belonged to a union (U.S. Bureau of Labor Statistics (2021)). Recently, however, there have been some glimmers of revival, including successful unionisation drives at Amazon and Starbucks that are raising questions about whether a resurrected union movement could help slow or reverse the trend toward increasing inequality. We potentially can learn something about that possibility by examining the impacts on inequality of the de-unionisation that has occurred over the last 50 years. De-unionisation can affect inequality through various channels: by shifting workers from the union sector, where both wages are higher and wage differentials among workers tend to be smaller, to the nonunion sector where those differentials are larger; through any impacts on employment (perhaps through a general lowering of labour costs); through any accompanying changes in inequality within the union sector; and through indirect effects on wage setting in nonunion firms. These latter, spillover, effects are important since their existence could imply that the impact of de-unionisation on inequality is larger than what is calculated based just on the shifting of workers from the union to the nonunion sector. In this paper, we build a model of the impact of unions on wage setting in the nonunion sector and use it in estimation based on Current Population Survey (CPS) data to re-assess the role of de-unionisation in movements in the wage structure in the US.

The idea that unions could impact nonunion wage setting goes back at least to [1]. The core idea raised in that paper, and discussed in subsequent papers such as [2], is that nonunion firms raise their wages in response to the ‘threat’ that their workers will unionise, presumably imposing extra costs beyond direct wage increases. Our model incorporates that ‘threat’ effect plus an added mechanism through which the union sector affects nonunion wages: a bargaining channel whereby the outside options of nonunion workers and, through that their bargained wage, are affected by their ability to find high paying union jobs. In a sense, both are threat channels, with one being the threat of workers to leave and find a union job and the other channel being the threat to unionise the nonunion workplace. In order to differentiate these two channels, we will call the effects due to concern about being unionised the traditional threat effect and the other channel the bargaining effect. Our model makes clear the difficulties inherent in separately identifying these two effects while controlling for selection into the union/nonunion sectors. Part of the contribution of this paper is to provide estimates of spillover effects through both of these channels. This has the potential to expand our notion of the extent of the impact of de-unionisation on the wage structure, with the goal of our empirical work being to estimate the size of any such expansion.

The existing literature on union wage spillovers implements a specification in which nonunion wages are regressed on the percent of organized workers in labour markets defined in various ways. Evidence based on variation in this proxy for union power is mixed and sensitive to the included control terms, with the preponderance of studies finding a small positive spillover effect, though [3] estimate a negative effect. In an important analysis, [4]...
carefully considers the role played by omitted variables, in turn introducing industry and state fixed effects. His analysis finds great sensitivity in estimates to the source of variation used, providing some context for the variation in estimates across earlier studies. When controlling for a wide range of potential omitted variables his results indicate at most a small positive effect of union power on nonunion wages.

In a recent paper, ? estimate the role played by union threat effects using variation in the unionisation rate at the industry×state level included as an additional covariate in their distribution regression approach. They find positive effects of the unionisation rate operating primarily at the part of the wage distribution just below the median and their counterfactual exercise indicates that spillovers double the measured impact of deunionisation in increasing wage inequality in the US. While this is very useful, this paper shares with all of the early analyses a lack of an identification strategy for addressing the potential endogeneity of the union proportion. This stems from the lack of an effective instrument for the union proportion.\(^3\) Virtually none of the papers in the literature even mention the twin problem of selectivity that could bias estimates: as the proportion of workers who are unionized declines, the composition of nonunion workers and firms will change.

In contrast to the existing literature on union spillovers that largely relies on reduced form estimation, our approach formalises union spillovers in a search and bargaining framework, endogenising the process of union formation and incorporating wage effects arising through differences in the bargaining process. In making clear what is being identified in the model and the variation used, we overcome the problems inherent in early studies of likely biases due to omitted characteristics and selection into the union sector, and we estimate an effect with a clear theoretical basis and interpretation.

Our model is based on that of ? (henceforth TD), whose work is informed by the contributions of ?, ?, and ?, among others. The TD model is centred around union threat effects which happen through the hiring channel. In the model, more skilled workers tend to dislike unionisation and firms skew their hiring toward these workers in order to stack the unionisation vote. Though this effect is certainly interesting, we believe it is likely of second order importance relative to a more direct firm response through raising wages to lessen the gains from unionising and direct union busting actions which raise the costs of unionisation. Our model focuses on these latter effects instead of the hiring channel.

Additionally, our framework is informed by ? (henceforth BGS), which formalises the impact of changing alternative job prospects (outside options) on wages due to changes in the industrial composition of work. Following BGS, we model local labour markets composed of industries and firms with workers able to transition between jobs in proportion to job

confirmed by ?. Conversely, ? finds a large positive effect, with a 10% increase in the proportion unionised associated with a 4% increase in nonunion wages. ? provides additional evidence in support of a positive spillover effect, while ? find a positive spillover effect at the industry level, but no significant effect operating in the local labour market. ? also find a positive correlation between the union proportion and wages, although ? finds this effect exists only for large establishments.

\(^3\) ? presents event studies of the enactment of Right to Work (RTW) laws in Idaho and Oklahoma, while noting that RTW's are more likely to be enacted in weak union states and, so, face endogeneity issues of their own. In an earlier version of the paper, (?), the authors extend the analysis of ? using variation in right-to-work laws in an event-study framework. They find evidence of reductions in nonunion wages with the introduction of right to work laws but the estimates are poorly defined because there are few states that switch RTW status in their time period.
prevalence. As in BGS we will exploit cross-city, within-industry variation - in our case, to identify the effect of declining unionisation and changes in the composition of union work on nonunion wages over the period 1980-2020.

Combining these elements, we derive an empirical specification which incorporates spillover effects operating through both the bargaining and standard threat channels, formalises selectivity, and makes it straightforward to see barriers to identification. Specifically, changes in outside options associated with the union sector may be correlated with unobserved local productivity shocks. As in BGS and ?, we overcome this problem using Bartik-style instruments related to worker outside options. For nonunion workers, outside options are related to the probability the worker could transit to a union job (which we allow to vary by industry and over time) times the expected wage the worker could get in that job. It also depends on expected wages in nonunion jobs in the local economy. The Bartik instruments use versions of these outside options based on start of period industry and union employment composition in a locality interacted with changes in industry growth, industry premia, and the probability of moving to a union job at the national level. It is this outside option for nonunion workers that identifies the bargaining channel for union effects. We get extra power to identify the bargaining effect because improvements in outside options have the same effect on bargained wages whether they stem from reduced probabilities of finding a union job or reductions in the number of high-rent nonunion jobs. That means we get identification from both changes in unionisation and shifts in industrial structure in both the nonunion and union sectors. We argue that the validity of our Bartik instrument depends on a random walk type assumption that we show implies an overidentifying restriction. We test that restriction and cannot reject it. We also show that within the context of the model, we identify the threat channel by the impact on nonunion wages of the interaction of the probability a firm in a given industry×city cell would face a union election (which shows the size of the direct threat) with the outside option value for union workers (which captures the size of what the firm needs to respond to in order to prevent unionisation). We construct and implement similar Bartik instruments related to this component.

The results from our estimation point to the importance of both spillover channels. Between 1980 and 2010, the mean real wage in the US fell 16% (holding composition in terms of education, experience, race and gender constant). Through a decomposition exercise based on our estimates, we find that de-unionisation accounts for 38% of the decline. A third of that impact arises from a standard shift share effect (because workers shifted away from higher paying union jobs) with the other two-thirds from spillover channels. Unions have spillover effects on nonunion wages and they are sizeable.

While both the traditional threat and bargaining effects show up significantly in our estimates, our decomposition exercise indicates that the spillover effects are almost entirely due to the latter. The threat probability was too low, even in 1980, to play a substantial role. This is potentially important for considerations about trying to raise wages through policy tools. The threat effect is unique – it can only be harnessed by increasing unionisation. The

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4 Although we specify the firm as the level at which workers become unionised in the model, our estimation exploits variation at the industry-city level. In that sense, firms can be thought of more accurately as establishments, or more generally, the level at which the unionisation vote takes place. Given the emphasis on the enterprise level in the Wagner Act, aggregation to the city level may ignore some complexity, particularly if there are notable transitions between different union/nonunion establishments in the same firm.
bargaining effect, in contrast, is more general. It is about more good paying jobs representing improved outside options for workers in all other jobs, something that ? and ? point out can have sizeable positive effects on wages in a location. Unions are one way to create such a higher wage option but other policies, such as eliminating non-compete arrangements could also have such an impact (?).

Although spillovers roughly tripled the standard shift share effect of unionisation over the long run, we find no evidence of bargaining spillovers in the 1980s – the decade of the largest declines in unionisation. This is because those declines were matched with increases in the union wage premium, increasing the value of the outside option of nonunion workers at the same time as declining probabilities of finding union jobs were reducing it. Our model provides an explanation for the increased wage premium in the 1980s that echoes an argument in ?. While both union and nonunion wages faced downward pressures from technological change, trade, etc., the substantial reduction in the risk of being unionised in the decade meant that, in addition, nonunion firms no longer had to pay higher wages in order to stave off unionisation. As a result, nonunion wages fell faster than union wages. After 1990, the threat of unionisation stabilized at a low level, causing the union wage premium to decline, and the outside option effect of unions began to reflect the falling unionisation rate alone. The potential lesson for any re-unionisation efforts is that spillover effects onto nonunion wages may arise through the traditional threat channel but the implied increase in nonunion wages will dampen the bargaining channel. Union jobs would be more plentiful but not pay as high a premium over nonunion jobs as before re-unionisation. Eventually, as the unionisation threat stabilized, the extent of spillover onto nonunion wages would increase, but that could take time to be fully realized.

Our work is also related to the substantial literature that investigates patterns in declining unionisation, estimating both movements in the union wage premium and the role of declining unionisation in driving increasing wage inequality. ? and ? provide comprehensive summaries of the research in this area following the early contribution of ?. ? provide the most comprehensive account of the relationship between union density and inequality in the U.S., introducing new survey data that allows them to push their analysis back to the 1930s. They find that increasing unionisation had a substantial impact on decreasing inequality after WWII while the reversal in the unionisation trend had a smaller effect on increasing inequality in the last 50 years. Their estimates allow for spillover effects onto nonunion wages but they do not study spillovers directly. Our results imply that spillovers may have played an important role in their estimated inequality impacts from unions and provide an explanation for why those impacts were less evident at the time of the big union decline in the 1980s.5

The remainder of the paper is organised as follows: In section 2, we present our versions of our model both with and without incorporated firm responses to unionisation threats. In

5Other papers in this literature include an important contribution by ?, who introduce a semi-parametric re-weighting technique building on the work of ? which attributes 14% of the increase in wage inequality over 1979-1988 (for men) to declining unionisation. Extensions of this work are found in ? and ?. Further studies by ?, ? and ? provide estimates of the union contribution to wage inequality, comparing patterns across male/female workers, private/public sector workers, and countries. In a recent contribution, ? compare the experience of Canada and the United States, focusing in particular in the increasing prominence of the public sector as a source of unionised employment. See also recent studies by ? and ?.
section 3, we describe the data and the construction of our key outside option variables and provide a discussion of our instrumental variables. Section 4 contains descriptive patterns and our estimation results. In section 5, we present a counterfactual exercise designed to demonstrate the impact of spillovers on movements in the wage structure and the role played by our two channels. Section 6 contains conclusions.

2 The Model

2.1 Model Set-up

Our model is based on that of ? (TD) which places union formation and wage setting in a search and bargaining model. Unions are able to bargain a higher wage because they can threaten to take the whole workforce out of production, while an individual, nonunion worker can only threaten to withdraw her own labour. In the TD model, firms employ workers of different skill levels who have different preferences about unions. In particular, since unions compress skill differentials, more skilled workers would vote against a union and less skilled workers would vote in favour. Firms facing a union threat can resist unionisation by taking advantage of these preferences: hiring a larger proportion of skilled workers in order to construct a workforce that will vote against a union. Our interest is in the potential spill-over effects of unions on nonunion wages and, partly for that reason, we focus instead on the concept of nonunion firms resisting unionisation through paying higher wages. We also view this type of wage emulation channel as well as direct legal and illegal campaigns by firms as likely more immediate responses to unionisation threats than altering the skill composition of the workforce. In addition, we will allow for the possibility of both direct and wage emulation effects of unions on wages in a given industry affecting wage setting in other industries in a local economy. Following ?(BGS), this can happen because having higher rent job options in a city increases the value of the outside option for workers in all industries as they bargain with their employers. Through the rest of the paper, we will refer to effects of unions on nonunion wage setting in order to resist unionisation as standard threat effects (to reflect that these are what have been discussed in the previous literature) and indirect effects through impacts on bargaining options as bargaining spillover effects. To focus attention on whether these channels are sizeable, we will alter the TD model by having only one skill level but multiple industries.

With that in mind, we start with an aggregate production function of the form:

\[ Q = \left[ \sum_i a_i Z_i^\chi \right]^{\frac{1}{\chi}}, \quad \chi < 1, \]

where \( Q \) is total output for the national economy and is equal to a CES aggregate of \( I \) intermediate goods indexed by \( i \). The intermediate goods \( (Z_i \text{'s}) \) are produced in local economies, or cities, of which there are \( C \). Thus, \( Z_i = \sum_c Y_{ic} \), where \( Y_{ic} \) is the amount of intermediate good \( i \) produced in city \( c \). The price for \( Q \) is normalized to 1 and the price for intermediate good \( i \) is given by \( r_i \). For simplicity, and to focus attention on wage setting, we will ignore firm entry and exit and assume that there is a fixed number of firms operating in each indus-
try by city cell. Thus, \(Y_{ic} = \sum_f y_{icf}\), where \(y_{icf}\) is the output of firm \(f\) operating in industry \(i\) in city \(c\).

Firms and workers operate in a labour market that includes frictions. Unemployed workers and vacancies posted by firms meet according to a matching function, \(M(U_c, V_c)\), where \(U_c\) is the number of unemployed workers in city \(c\) and \(V_c\) is the number of unfilled vacancies. As is standard in the search literature, we will assume that the matching function is constant returns to scale. BGS show that in steady state this implies that the probability that a vacancy in city \(c\) meets an unemployed worker, \(q_{vc}\), and the probability that an unemployed worker in city \(c\) meets a vacancy, \(q_{uc}\), can be written as functions of the employment rate for the city, \(ER_c\). We will assume that workers are not mobile across cities and that there is no on-the-job search.\(^6\) Our model is partial equilibrium in that we treat \(q_{vc}\), \(q_{uc}\), and \(ER_c\) as given rather than solving for them from the model. We make this simplifying assumption in order to maintain our focus on wage outcomes.

Once workers and firms meet, they bargain a wage to divide the match surplus. Following TD, in a nonunion firm, this bargaining is between the individual worker and the firm while in a union firm it is between the set of employees and the firm. Employees at nonunion firms have the opportunity each period to vote on whether to form a union. Once unionised, a firm stays unionised – there is no decertification. Both union and nonunion firms choose the optimal number of workers to hire, taking account of the bargained wage.

### 2.2 Workers

We begin by characterizing the choices and environment faced by workers. We will work in discrete time and assume that all workers have the same skill level. At any moment, a worker can be unemployed and searching for work or employed at one of three types of firms in a particular industry. The first type of firm is a simple nonunion firm in which the firm and worker bargain a wage and the firm chooses an optimal number of employees without direct regard to the threat of unionization. As we will see, these firms exist in situations where the costs and benefits of unionization are such that workers at the firm do not want to form a union. Workers in these firms have a value function given by:

\[
W_{icf}^n(w_{icf}^n) = w_{icf}^n + \rho(\delta U_{ic}^n + (1 - \delta)W_{icf}^n(w')),
\]

where \(\rho\) is the discount rate, \(w_{icf}^n\) is the nonunion wage that would be paid at firm \(f\) in the given \(i-c\) cell, \(\delta\) is an exogenous probability that the worker’s job is terminated, and \(U_{ic}^n\) is the value of unemployment in city \(c\) for a worker whose previous job was a nonunion job in industry \(i\). In this specification, \(w'\) corresponds to the wage that will be paid in the next period if the job is not terminated. Following TD, we assume that workers and firms believe that next period’s wage will be set optimally and that they cannot affect it through actions this period. This assumption rules out, for example, the relevance of reputations.

Alternatively, workers may vote to unionise, in which case their value function is:

\[
W_{icf}^u(w_{icf}^u) = w_{icf}^u + \psi_{icf} + \rho(\delta U_{ic}^u + (1 - \delta)W_{icf}^u(w')),
\]

\(^6\) provide a model with mobility.
where $u$ superscripts correspond to unionised values, and $\psi_{icf}$ is the non-wage benefit to workers from being in a union in this particular firm, which could be related to the work environment in the firm. We assume that a value for $\psi_{icf}$ is drawn separately for each firm from a common distribution, $f(\psi)$, and that the union amenities are created by the union itself rather than the firm and do not enter the cost function for the firm.

Firms can respond to a union threat by paying workers a wage that is high enough to make them indifferent between unionising and remaining non-union (what we will call an emulation wage). Given this, there is a third possible type of firm which we will call a union emulating firm in which workers are paid a wage, $w^{n*}_{icf} > w^n_{icf}$ but do not get union benefits. As a result, the value function associated with working at this type of firm is the same as for the simple non-union firm but with a wage of $w^{n*}_{icf}$.

The other potential state for workers is being an unemployed searcher. For reasons that will become apparent momentarily, the value of unemployment differs by previous industry and by whether the person was a union or nonunion worker in their previous job. The value function for a person who was formerly in a nonunion job in industry $j$ is given by the following equation (an analogous expression exists for those formerly in a union job in industry $j$):

$$U^n_{jc} = b + \rho \left[ q_{uc} \left( T^n_{jc} \sum_i \eta^{u}_{ic} W^{u}_{ic} + (1 - T^n_{jc}) \sum_i \eta^{n}_{ic} (p^n_{ic} W^{n}_{ic} + (1 - p^n_{ic}) W^{n*}_{ic}) \right) + (1 - q_{uc}) U^n_{jc} \right].$$

where $b$ is the flow value of being unemployed; $\eta^{u}_{ic}$ and $\eta^{n}_{ic}$, respectively, are the proportions of employment in the union and nonunion sectors in city $c$ that are in industry $i$. For workers in city $c$ and previously employed in a nonunion job in industry $j$, the probability of finding a job in the union sector is $T^n_{jc}$. With probability, $(1 - T^n_{jc})$, these workers match with a nonunion firm. We allow the probability of finding a union job to differ by whether the worker was previously in a union job and by their industry of previous employment.\(^7\) This is the reason for indexing the value of unemployment by previous industry and union status.

We assume that search is random – that, for example, workers are unable to differentiate between nonunion firms paying regular nonunion wages and nonunion firms paying emulation wages.\(^8\) As a result, they find employment in proportion to the relative abundance of these firms as measured by $p_{ic}$, which is the fraction of jobs in industry $i$ in city $c$ that are in non-emulation firms. $W^{u}_{ic} = E(W^{u}_{icf})$ is the expected value of employment in a union firm in the $i - c$ cell with the expectation taken across the distribution of union firms in that cell (the other expectations are defined analogously).

Taken together, (3) says that an unemployed worker gets a flow value, $b$, and with probability $q_{uc}$ meets a vacancy. Workers have different propensities for finding employment

\(^7\)For union job searchers the probability is $T^n_{jc}$.

\(^8\)Our model therefore abstracts away from issues related to workers queueing for union jobs (see \(^?\) for a theoretical treatment of queuing and supportive empirical evidence). This queuing mechanism could imply an additional spillover channel whereby the existence of union firms drives down vacancy filling rates in the nonunion sector, pushing up wages. The prevalence of queueing is likely driven by union wage premia and the relative likelihood of finding union work such that queueing effects are likely to enter through the outside option channel in our model.
in the union and nonunion sectors, but otherwise find jobs in proportion to their share of local employment. The expression implies that the value of search is higher when the local employment structure has more high value industries and more unionised firms since that means searchers are more likely to find those high value jobs. To simplify other expressions in the model we will often summarize the expected value of a job as:

\[
\bar{W}_{jc}^n = T^n_{jc} \sum_i \eta^n_{ic} W^n_{ic} + (1 - T^n_{jc}) \sum_i \eta^n_{ic} \left( p^n_{ic} W^n_{ic} + (1 - p^n_{ic}) W^n_{ic}^* \right).
\] (4)

2.3 Firms

All firms in a given industry-city cell have a common production function given by:

\[
y_{icf}(n) = \epsilon_{ic} n_f - \frac{1}{2} \sigma_i n_f^2,
\]

where \( \epsilon_{ic} \) is a local productivity shock, \( n_f \) is the number of employees, and \( \sigma_i > 0 \) is a parameter reflecting potential span of control issues.\(^9\) This specification implies that technology is common across cities within an industry but that there is comparative advantage in producing each intermediate good by city. We assume that the technology is common to all three types of firms (unionized, non-union, and union emulators). The literature on union effects on productivity seems to us to be inconclusive and so we adopt an agnostic take in which unions affect firm activity by affecting wages but not through technological adaptations.\(^10\) We assume that the \( \sigma_i \)'s are sufficiently smaller than 1 such that, combined with the assumption of a fixed number of firms in each \( i-c \) cell, they imply that production of any good is spread across cities.\(^11\)

Firms choose a number of vacancies to post in order to attain a profit maximizing number of employees given the bargained wage. The cost of hiring is linear in the number of vacancies posted. As a result, the value function for a nonunion firm is given by:

\[
J^n_{icf}(n) = \max_v r_i y_{icf}(n') - w^n_{icf}(n') n' - \kappa v + \rho J^n_{icf}(n'),
\]

subject to the equation of motion:

\[
n' = n(1 - \delta) + q_{vc} v,
\]

where \( J^n_{icf}(n) \) is the value of non-union firm \( f \) when it ends the previous period with \( n \) employees. Of those, a fraction, \( \delta \), leave the firm for exogenous reasons. The firm posts \( v \) vacancies at a cost, \( \kappa \), per vacancy and fills them with probability, \( q_{vc} \). Following TD, we will assume that the firm can only post positive vacancies and that the number of vacancies

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\(^9\)For notational simplicity, we will drop the \( f \) subscript on \( n_f \).

\(^10\) and ? summarise the early research in this area which finds largely inconclusive and mixed evidence on the effect of unionisation on productivity.

\(^11\) We work with a quadratic production function to permit tractability in deriving our wage expressions. It is worth noting that TD showed that using a Cobb-Douglas type form for production has the unfortunate implication that general productivity shifts such as \( \epsilon_{ic} \) do not determine wages. In that sense, our results are not perfectly generalizable but a quadratic function captures the main points we want to emphasize.
is sufficiently large that we can treat $q_{vc}$ as a deterministic number according to a law of large numbers. This allows us to rewrite the value function as:

$$J_{icf}^n(n) = \max_{n'} r_i y_{icf}(n') - w_{icf}^n(n') n' - \kappa n' - n(1 - \delta) q_{vc} + \rho J_{icf}^n(n').$$

Note that we have written the value function as if the firm exists in a stationary environment in which it assigns a probability of zero to its workers trying to unionise in the future. We will return to that assumption below. The value functions for firms when they are unionised or acting as union emulators are identical in structure to the nonunion firm value function, with the only change being the substitution of the relevant wage ($w_{icf}^{n*}(n')$ or $w_{icf}^u(n')$) in the expression. We next turn to using these value functions in combination with those for the workers to characterize wage bargaining solutions. Once we have those solutions, we will be in a position to discuss union formation.

### 2.4 Wage Determination

#### 2.4.1 Collective Bargaining

Once the firm hires workers, the workforce will vote on unionisation. If a union is formed then the union bargains collectively on behalf of workers. Following TD, wages are bargained according to Nash bargaining over the entire surplus to production from hiring $n$ workers. By bargaining collectively, the union is able to effectively threaten that the entire workforce will quit this period. Such an action would impose two costs on firms: first, the firm would produce zero units of output this period; second, as the firm employs $n$ workers every period in steady state, the number of vacancies required to achieve this optimal workforce will be much larger, increasing vacancy filling costs. These costs are taken into account when determining firm and worker surplus.

Wages (in a unionised firm) are set according to the Nash Bargaining condition:

$$\beta S^u = (1 - \beta)n(W_{icf}^u(w) - U_{ic}^u),$$

where $S^u$ represents the firm’s surplus. On the right hand side is the sum of workers’ surplus, which is given by the gain to employment for all workers hired by the firm. Since the workers are identical, we use a specification that focuses on the total surplus and assume that the union members will all get an equal share of the part of the surplus captured by the union. This ignores issues related to seniority, for example.\(^{12}\)

We will focus on a steady state in which the wage and optimal number of workers for a firm are constant across periods. Given this, the workers’ surplus can be re-written as:

$$W_{icf}^u(w) - U_{ic}^u = \frac{1}{1 - \rho(1 - \delta)}(w_{icf}^u(n_{icf}^u) + \psi_{icf}) - \frac{(\rho - 1)b}{(1 - \rho(1 - \delta))(1 - \rho(1 - q_{uc}))} \tilde{W}_{icf}^u.$$

\(^{12}\)See ?? who present evidence of the importance of seniority for layoffs and promotions, and see ? for evidence that the seniority wage profile is steeper under collective bargaining.
On the firm side, the surplus from a successful bargain with a union is given by the difference between producing this period with \( n_u \) workers (the optimal number of workers with a bargained union wage) and not producing this period along with the cost of rehiring the entire workforce the next period. The firm surplus can be expressed as follows (with a detailed derivation provided in Appendix A):

\[
S^u = (r_i y_{icf}(n_{icf}^u) - n_{icf}^u w_{icf}(n_{icf}^u)) + \rho(1 - \delta)\frac{k n_{icf}^u}{q_{vc}}.
\]

The surplus equals the current period profit plus the cost of replacing the portion of the workforce that would not normally turn over, which is the relative cost of being in the bargaining break-down option.

Solving the bargaining expression for steady state wages yields:

\[
w_{icf}(n_{icf}^u) = \frac{\beta(1 - \rho(1 - \delta)) r_i y_{icf}(n_{icf}^u)}{(1 - \beta \rho(1 - \delta)) n_{icf}^u} + \frac{\beta \rho(1 - \delta)(1 - \rho(1 - \delta)) \kappa}{(1 - \beta \rho(1 - \delta)) q_{vc}}\]

\[
+ \frac{(1 - \beta)(1 - \rho)}{(1 - \beta \rho(1 - \delta))(1 - \rho(1 - q_{uc}))} b - \frac{(1 - \beta)}{(1 - \beta \rho(1 - \delta))} \psi_{icf} \]

\[
+ \frac{(1 - \rho) \rho (1 - \beta) q_{uc}}{(1 - \beta \rho(1 - \delta))(1 - \rho(1 - q_{uc}))} \tilde{W}_{ic}^{n}.
\]

### 2.4.2 Individual Bargaining

Non-union wages are also set through Nash bargaining. However, the value for the firm of the option corresponding to a break down in negotiations relates only to the loss of an individual worker (and the indirect effects the removal of one worker has on the others) rather than to the complete stoppage of production that occurs under collective bargaining. The Nash bargaining condition is the same as before, and we derive the following expression for the firm surplus (see Appendix A.1 for details):

\[
S^n = r_i \frac{\partial y_{icf}(n_{icf}^n)}{\partial n} - w_{icf}(n_{icf}^n) \frac{\partial w_{icf}(n_{icf}^n)}{\partial n} + \rho(1 - \delta)\frac{\kappa}{q_{vc}}.
\]

Substituting this expression into the Nash bargaining condition and (again, following TD) solving the differential equation in wages yields the following expression for non-union wages:

\[
w_{icf}(n_{icf}^n) = \frac{1 - \rho(1 - \delta)}{1 - \beta \rho(1 - \delta)} \frac{\beta r_i}{1 + \beta} \left( \frac{\partial y_{icf}(n_{icf}^n)}{\partial n} + \beta \epsilon_{ic} \right) + \frac{\beta \rho(1 - \delta)(1 - \rho(1 - \delta)) \kappa}{(1 - \beta \rho(1 - \delta)) q_{vc}}
\]

\[
+ \frac{(1 - \beta)(1 - \rho)}{(1 - \beta \rho(1 - \delta))(1 - \rho(1 - q_{uc}))} b + \frac{(1 - \rho) \rho (1 - \beta) q_{uc}}{(1 - \beta \rho(1 - \delta))(1 - \rho(1 - q_{uc}))} \tilde{W}_{ic}^{n}.
\]

This wage differs from the union wage in two ways. First, under collective bargaining the union negotiates over the surplus generated from total output while for non-union workers what matters is the marginal surplus. As a result, individual wages are determined by average output for union workers but marginal product for nonunion workers. Second, union wages
include a compensating differential component with wages declining when union amenities are greater. On the other hand, outside options – which are determined by both the value of non-work time and the expected value of finding a new job after unemployment – have the same effects on the union and nonunion wages.

2.4.3 Wage Equations: Empirical Specifications

The wage equations derived to this point include firm size as an argument of either the average or marginal product of workers. In Appendix A.3, we derive the optimal firm size expressions for union and nonunion firms and then substitute those into the wage equation. Thus, we implement quasi-reduced form wage specifications that allow us to avoid dealing with firm size endogeneity issues and with problems with the CPS data where there is categorical data on the size of the whole firm rather than on the establishment or workplace. It is the latter that is most relevant for our analyses.\textsuperscript{13} It is worth noting that nonunion firms have a larger optimal size, though this is mitigated to some extent by the fact the union amenities lower the wage differential between union and nonunion firms.

In our empirical work, we will focus on linearizations of wages around the point where all cities have the same proportions in each industry.\textsuperscript{14} This is done partly to make explicit the role of the employment rate in the city, $ER_c$. In steady state, the worker job finding rate, $q_{uc}$ and the firm worker finding rate, $q_{vc}$ can be written as functions of $ER_c$. The other main driving force is the outside option for workers, which is a function of the expected value of a job ($\tilde{W}^n_{ic}$ for a nonunion worker currently in sector $i$ in city $c$). Working from the fact that the expected value of a job is a function of wages, one can show that the outside option for a nonunion worker can be represented as a weighted average over wages in different sectors in the local economy. We will write that weighted average as $E^n_{ic}(w)$ for a nonunion worker and $E^u_{ic}(w)$ for a union worker.

Thus, wages can be written as:

$$w^n_{icf} = \gamma^n_{0i} + \gamma^n_{1i} E^n_{ic} + \gamma^n_{2i} ER_c + \gamma^n_{4i} \epsilon_{ic} \tag{5}$$

and

$$w^u_{icf} = \gamma^u_{0i} + \gamma^u_{1i} E^u_{ic} + \gamma^u_{2i} ER_c - \gamma^u_{3i} \psi_{icf} + \gamma^u_{4i} \epsilon_{ic} \tag{6}$$

for nonunion and union workers, respectively, where

$$E^n_{ic} = T^n_{ic} \sum_j \eta^n_{jc} w^n_{jcf} + (1 - T^n_{ic}) \sum_j \eta^n_{jc} \left( p^n_{jcf} w^n_{jcf} + (1 - p^n_{jcf}) w^n_{jcf} \right) \tag{7}$$

and

$$E^u_{ic} = T^u_{ic} \sum_j \eta^u_{jc} w^u_{jcf} + (1 - T^u_{ic}) \sum_j \eta^u_{jc} \left( p^u_{jcf} w^u_{jcf} + (1 - p^u_{jcf}) w^u_{jcf} \right), \tag{8}$$

where $p^n_{jcf} w^n_{jcf} + (1 - p^n_{jcf}) w^n_{jcf}$ is the observed mean nonunion wage. Note that the union and nonunion wage equations have different error terms, with the nonunion error term consisting

\textsuperscript{13} Although we have used ‘firms’ throughout this paper as our terminology of choice, one can think of union certification at the level of the workplace or establishment without making any modification to the model.

\textsuperscript{14} See Appendix A.4 for details.
of the productivity shock, $\epsilon_{ic}$, while the union error term includes the productivity shock but also the unobserved (to the econometrician) value of union amenities, $\psi_{icf}$. Given that we have assumed, so far, that workers are identical, higher wages in an industry correspond to rents – differences in pay over and above what is required for the marginal worker to want to join that industry. Those rents are maintained because of the frictions in the labour market. It is important that we are considering rents since wage differences across industries that correspond to compensating differentials (say, for dangerous work) cannot be the basis of bargaining a higher wage with your current employer. If a higher wage corresponds to a compensating differential then there is no net gain to moving to the dangerous job and, so, no basis on which to threaten your current employer during bargaining.

### 2.5 Union Determination

Our theoretical analysis occurs at the level of the firm and, therefore, we are interested in the question of whether firms become unionised. We assume there is no decertification, so unionisation is an absorbing state for a firm.\textsuperscript{15} But all firms – union and nonunion – die with probability $\delta_d$ in a period (which is a component of the separation rate, $\delta$, seen earlier). The dying firms are replaced with new firms started by new entrepreneurs. In steady state, the number of dying firms equals the number of newly born firms. All firms are born nonunion and then workers at the firm decide whether to unionise. Recall that firms are born with a draw of a level of amenities that would be provided at that firm if it were unionised, $\psi_{icf}$.

#### 2.5.1 Union Determination Without Firm Responses

We begin with a simple model in which firms are passive players in unionisation. That is, unionisation is determined entirely by the workers and firms do not try to respond by, for example, emulating union wages. In terms of the model derived so far, this implies that the probability of meeting an emulating firm $(1 - p_{ic}^*)$ is zero and the outside option wage terms are adjusted accordingly.

In the simple model, the workers at a firm compare the value of the job continuing as a nonunion job to the value of the job being union minus the cost of unionising. The wages they use in this exercise are the ones we arrived at in the previous section that reflect the optimal hiring decisions by firms. We will assume that the decision is made according to a median voter model with the median voter not at risk of losing her job when employment is reduced after unionisation. We also assume that workers do not care about the employment outcomes of those who do lose their jobs, implying that we can focus exclusively on wages. Given this, a firm becomes unionised if:

$$W_{ucf}(w_{ucf}) - \lambda^*_{ct} > W_{ncf}(w_{ncf}),$$

where $\lambda^*_{ct}$ is the fixed cost to workers of unionising a firm in a city $c$. Substituting in steady

\textsuperscript{15}Decertification is much less common than certification. Using election data from the NLRB (discussed below) we find that certification elections outnumber decertification elections over 1980-2010 by at least 5 to 1. The same is true if we consider the number of workers involved in elections. \footnote{present a figure showing the ratio of eligible workers certified over the time horizon 1978-2017.}
state expressions for the value functions based on (2) and (1), we can define an index function:

\[ I_{icf} = (w^u_{icf} - w^n_{icf}) + \psi_{icf} - (1 - \rho(1 - \delta))\lambda^*_{ct}, \] (9)

such that a firm is unionised if \( I_{icf} > 0 \) and remains nonunion otherwise. Here, we have assumed that all workers in the firm share the costs of unionisation equally. The term multiplying the fixed cost per worker puts the one-time fixed cost of unionising in the same present value terms as the flow of wages and union amenities.

We can substitute in the union and nonunion linearized wage expressions, (6) and (5), into (9) to obtain:

\[ I_{icf} = \alpha_0 + \gamma^u E^n_{ic} - \gamma^n E^n_{ic} + \alpha_2 ER_c + (1 - \gamma^u_3)\psi_{icf} - \lambda_{ct} + \alpha_4 \epsilon_{ic}, \] (10)

where \( \lambda_{ct} = (1 - \rho(1 - \delta))\lambda^*_{ct} \) and \( \alpha_4 \geq 0 \).\(^{16}\)

This is a very standard selection set-up and implies:

\[ E(w^n_{icf}|I_{icf} \leq 0) = \gamma^n_0 + \gamma^n_1 E^n_{ic} + \gamma^n_2 ER_c + \gamma^n_4 E(\epsilon_{ic}|I_{icf} \leq 0). \] (11)

Notice that the error-mean term is a function of \( \lambda_{ct} \) – the cost of unionising – but the unconditional nonunion wage equation is not. Thus, measures of the cost of unionising are available as exclusion restrictions that identify selection effects separately from direct determinants of the nonunion wage. If we consider two cities, \( c \) and \( c' \), that are identical except that \( c' \) has higher costs of unionisation then unionisation will be lower in \( c' \). Moreover, because the coefficient on the productivity term, \( \epsilon_{ic} \), is positive in the index function and recalling that \( \psi_{icf} \) and \( \epsilon_{ic} \) are assumed to be independent, union firms tend to have higher productivity. Thus, the marginal firms that would be unionised in \( c \) but non-union in \( c' \) will be at the low end of the productivity range for union firms but the high end for nonunion firms. This has implications for a simple specification using the proportion union to capture spillover effects, as the estimated coefficient on the union proportion would be biased downward. Industry-city cells with higher unionisation rates would be ones with lower productivity among nonunion firms.

In what follows, we first estimate the regression specification given by (11). This allows us to investigate the presence and size of spillovers of union power on nonunion wages. We do not attempt to estimate a specification for local union wages because low unionisation rates, especially in the later years of our sample, imply sample sizes in industry × city cells that are too small to work with.

### 2.5.2 Incorporating Firm Responses

Next, we consider a more realistic setting in which firms at risk of being unionised can respond to forestall unionisation. In TD, firms respond by hiring more skilled workers who they know will vote against unionisation. While this response is possible, it seems to us to likely be of second order importance relative to more direct responses. In real world

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\(^{16}\)This condition on \( \alpha_4 \) is shown in Appendix A.4 and arises because nonunion wages are determined by the marginal product of one worker while union wages relate to average product of all workers.
descriptions of firm reactions to the threat of unionisation,\textsuperscript{17} firms adopt some combination of three possible responses: 1) paying higher wages to reduce the net benefit of unionising; 2) providing better working conditions to match what workers would get in a union setting (in the context of our model where only union workers get the amenities, $\psi_{icf}$, nonunion firms would provide similar amenities to their workers); and 3) intimidation (which, in our model, would correspond to increasing the cost of unionisation, $\lambda_c$). We begin by focusing on the wage emulation response, returning to the other potential responses later.

Before discussing the responses, note that firms do not need to consider employing any response if their workers are happily nonunion. That is, if the costs of unionising, the amenities that would be available to the workers at this firm if they unionise, the union wage they would get if they organise, and the nonunion wage they get if they don’t are such that $I_{icf} < 0$ then there is no reason for the firm to bear costs to incentivize its workers not to form a union. We are interested in the set of firms that are unionised (i.e., for which $I_{icf} > 0$) but near the margin of being so in our first, non-responsive firm model. Recall that the fixed cost of forming a union depends on the legal climate in each state while the non-wage benefits of unionising are firm specific. As a result, not all firms in the same industry and city will have the same worker preferences about unionisation.

\section*{Wage Response}

First, consider the possibility that firms respond to their workers’ desire to unionise by offering the workers a wage that just offsets the benefit of unionising. In particular, based on the discussion of worker preferences on whether to unionise underlying the index function (9), the wage the firm would need to offer to make workers indifferent about forming a union is:

$$w^*_{icf} = w^u_{icf} - \lambda_c + \psi_{icf}.$$  \textsuperscript{17}

Now consider worker preferences about unionising when they take account of the non-emulation nonunion wage we derived earlier, $w^o_{icf}$, the cost of unionising and the union wage. In this situation, we can define a $\psi^*_{icf} = \lambda_c - (w^u_{icf} - w^o_{icf})$: the value for amenities such that workers at firm $f$ are indifferent between whether they organise or not. For $\psi \leq \psi^*_{icf}$, workers will not organise, and the firm will pay the regular nonunion wage, $w^o_{icf}$. For $\psi > \psi^*$, workers will prefer to unionise if offered the regular nonunion wage and firms will consider the value of offering the emulation wage, $w^*_{icf}$, instead and forestalling unionisation. We show in Appendix A.5 that the firm will be willing to pay the higher, emulation wage until the point where the costs and benefits to unionisation overlap, which happens when $\psi_{icf} = \lambda_c$. At that point, the emulation wage equals the union wage. Beyond it, the wage required to prevent union formation exceeds the union wage, and to fight the union would lower firm profits. For all $\psi_{icf} > \lambda_c$, then, the firm will become unionised. Thus, we can characterize the firm union status as follows:

- $\psi_{icf} < \psi^*_{icf}$: $f$ is nonunion and pays $w^o_{icf}$,
- $\psi^*_{icf} \leq \psi_{icf} < \lambda_c$: $f$ is nonunion but emulates unionised firms and pays $w^*_{icf}$,
- $\psi_{icf} > \lambda_c$: $f$ is unionised.

\textsuperscript{17}See The Guardian: ‘Pay a living wage’: Bernie Sanders accuses Disney of dodging fair pay, The Guardian: Why Target’s anti-union video is no joke, and The Guardian: Delta workers seeking to unionize say they are ‘under siege’ by management.
\[ \psi_{icf} \geq \lambda_c : f \text{ is union and pays } w_{icf}^u. \]

Note that both cut-offs rise with \( \lambda_c \), so that states that implement policies that raise the cost of unionising will have more nonunion firms. As the unionising costs rise, the actual nonunion wage will not change. However, the observed nonunion wage will decline because the fraction of nonunion workers who are paid the higher, emulation wage, will decline. Moreover, the emulation wage \( w_{icf}^* \), will also decline. Thus, we would observe a decline in unionisation combined with an increase in the observed union wage differential. This pattern might seem to imply that de-unionisation is arising because of union rigidity on wages. But this can happen even without unions being rigid (i.e., even though union wages will decline with declines in \( \epsilon \)). The pattern of declining unionisation with an increasing union wage differential is what we observe during the 1980s.

To obtain an empirical specification related to the model including union emulation, first note that the linearized version of the union emulation wage is given by:

\[
\begin{align*}
w_{icf}^* &= \gamma_u^0 + \gamma_u^1 \tilde{E}_{icf}^u + \gamma_u^2 E_{icf} + (1 - \gamma_u^3) \psi_{icf} - \lambda_{ct} + \gamma_u^4 \epsilon_{icf}
\end{align*}
\]

There is one additional complication that we have yet to address, which is that for certain values of \( \epsilon \) there may be no range over which workers wish to unionise. In terms of the thresholds defined above this occurs when \( \psi^* = \lambda_c - (w^u - w^n) \geq \lambda_c \), which occurs when the nonunion wage exceeds the union wage. In this instance, for all values of \( \psi \), workers will prefer to remain nonunionised. We define the productivity draw that equalises the costs and benefits of unionisation as \( \epsilon^* \). For productivity draws above this threshold the union wage will exceed the nonunion wage and there will be some set of \( \psi \) values over which unionisation is preferred.

The observed mean nonunion wage in an \( ic \) cell is given by a weighted average of the actual nonunion wage, and the emulation wage. We define this observed wage as \( \bar{w}_{ic}^n \):

\[
E(\bar{w}_{ic}^n) = \frac{Pr(I_{icf} < 0)}{Pr(I_{icf} < 0) + Pr(I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*)} E(w_{icf}^u | I_{icf} < 0) + \frac{Pr(I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*)}{Pr(I_{icf} < 0) + Pr(I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*)} E(w_{icf}^* | I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*)
\]

The weights are the probability of firms being of each type conditional on being observed as a nonunion firm. To form this regression equation, we first need to specify the probabilities that make up the weights:

\[
Pr(I_{icf} < 0) = \int_{-\infty}^{\psi_{icf}^u} \frac{\lambda_c - \Delta_{icf}}{1 - \gamma_u^3} f(\psi) g(\epsilon) d\psi d\epsilon
\]

and

\[
Pr(I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*) = \int_{\epsilon^*}^{\infty} \frac{\lambda_c}{1 - \gamma_u^3} f(\psi) g(\epsilon) d\psi d\epsilon,
\]

18 To see this note that the cut-off for the lower end of the emulation range, \( \psi_u = \lambda_{ct} - (w^u - w^n) \), will rise faster than the upper end (which is \( \lambda_{ct} \)) as \( \lambda \) increases. This can be shown by noting that \( \frac{\partial \psi_u}{\partial \lambda_{ct}} = 1 - \frac{\partial w^u}{\partial \lambda_{ct}} \) and that \( \frac{\partial w^u}{\partial \lambda_{ct}} < 0 \) since union wages are lower when union related amenities are higher.
where $\Delta = \alpha_0 + \gamma^n_1 E^n_{ic}(w) - \gamma^n_1 E^n_{ic}(w) + \alpha_2 ER_c$, the union benefit threshold is $\psi^* = \frac{\lambda_c - \Delta - \alpha_2}{(1 - \gamma_3)}$, and recalling that the range for the productivity shock, $\epsilon_{ic}$ is $[0, \infty]$. The first probability is the probability that workers would choose to be nonunion and the second is the probability that firms are nonunion but emulate union wages.

Substituting equation (11) for $E(w^n_{icf} | I_{icf} < 0)$ and the relevant expectation over (12) for $E(w^n_{icf} | I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*)$ in equation (13), we arrive at an expression for the expected nonunion wage that incorporates firm emulation responses:

$$E(\bar{w}_{ic}) = \gamma^n_{0i} + \gamma^n_1 (1 - P(\psi_{icf}, \lambda_c)) E^n_{ic} + \gamma^n_1 P(\psi_{icf}, \lambda_c) E^n_{ic} + \gamma^n_2 ER_c + \mu(\epsilon_{ic}, \psi_{icf}, \lambda_c),$$

where $P(\psi_{icf}, \lambda_c)$ is the probability of a nonunion firm being a wage emulator given in (15).¹⁹

We implement both the specification assuming no firm responses, equation (11), and the one incorporating wage emulation, equation (16), in the empirical work that follows. Note that the emulation specification differs from the no-response specification, in part, because of the inclusion of the value of outside options for union workers. The latter is based on the probability of a union worker finding a union job and should not determine nonunion wages in the absence of the standard threat effect. In a separate estimation not reported here, we estimated a linearized version of (16) with $E^n_{ic}$, $E^n_{ic}$, $P(\psi_{icf}, \lambda_c)$, and $ER_c$ as regressors. The fact that $E^n_{ic}$ entered significantly in that estimation is strong evidence in favour of the standard threat effect being real.

**Amenity and Intimidation Responses** The other ways firms can respond to a unionisation threat are to provide amenities that match what unions provide and to raise the costs of unionising. We discuss both in more detail in Appendix A.5. For amenities, we describe a situation in which providing amenities has an increasing and convex marginal cost function when the firm chooses the wage emulation response. We show that, intimidation is a more cost effective response when workers are close to indifferent about unionising but becomes less effective when the level of intimidation required is large (i.e., as the value of $\psi$ rises). We show that under reasonable assumptions there is a new cut-off value, $\psi^*_{icf}$.

---

¹⁹$\mu(\epsilon_{ic}, \psi_{icf}, \lambda_c)$ is an expected error mean term equal to $\gamma^n_4 E(\epsilon_{ic} | I_{icf} < 0) - \lambda_c + E(\gamma^n_4 \epsilon_{ic} + (1 - \gamma^n_3) \psi_{icf} | I_{icf} > 0, \psi_{icf} < \lambda_c, \epsilon > \epsilon^*)$. 

16
such that, $\psi_{icf}^* < \psi_{icf}^b < \lambda_c$. Firms with $\psi_{icf} < \psi_{icf}^b \leq \psi_{icf}^b$ will use intimidation and, as a result, will remain nonunion and pay the nonunion wage. Firms with $\psi_{icf}^b < \psi_{icf} \leq \lambda_c$ will be nonunion but pay the emulation wage. As before, firms with $\psi_{icf}$ below $\psi_{icf}^*$ will be nonunion, paying the nonunion wage, and firms with $\psi_{icf}$ above $\lambda_c$ will be unionised. Thus, introducing intimidation serves to expand the region over which unions are nonunion and pay the simple nonunion wage to include the range over which firms use intimidation. We do not have a way to separately identify $\psi_{icf}^b$ from $\psi_{icf}^*$ and so cannot distinguish between a model in which there is no intimidation and the relevant cut-off for the nonunion region is $\psi_{icf}^*$ and one in which there is intimidation and the relevant cut-off is $\psi_{icf}^b$. Since both cut-offs are functions of the same variables – $\lambda_c$, the expected rents, and the employment rate – there is essentially no impact on our empirical specifications of including or not including intimidation. We will proceed as if there is no intimidation in order to simplify the exposition.

3 Implementation of the Wage Specifications

In this section we present details and results regarding estimation of the nonunion wage equations (11) and (16) derived from the model presented in Section 2.5.

3.1 Data


From this data, we keep all workers between the ages of 25-65 who do not report being in school either full-time or part-time. We follow in the construction of our wage data, working with weekly wages. We use an aggregated grouping of industry codes based on the 1980 industrial classification from the Census Bureau. We obtain a consistent industry classification using crosswalks provided by IPUMS and the Census Bureau that map the 1970, 1990, and 2000 industry codes to the 1980 classification. The result is a consistent classification system with 51 industries. Appendix B contains additional processing details.

We construct a set of cities with as consistent geographic boundaries as possible given data limitations in the CPS. We are constrained by the number of SMSA’s available in the May extract data and end up with 43 cities. Making use of the limited number of counties identified in the CPS, we are able to create a set of cities which are reasonably, though not always perfectly, consistent over time. The final geographic definition we use pools data for these 43 cities and the remaining population. Specifically, we create additional regions made up of the remaining state population absent the population living in these 43 cities.

20The metropolitan area definition used by the IPUMS identifies a general pattern of expanding metropolitan area definitions over time that we overcome to some extent, but not perfectly: https://usa.ipums.org/usa/vollii/county_comp2b.shtml. Estimation using states as the geographic unit yields very similar results, suggesting that issues related to geographic definitions are not driving our results.
In the end, our core geographic measure is composed of 93 areas that are fairly consistently defined over the course of the sample period.

Additionally, we use data on union elections to proxy for the costs of unionisation, $\lambda_{ct}$, in our model. The idea is that locations where the proportion of union certification elections that result in a certification is high are more union friendly. To obtain these proportions we use National Labor Relations Review Board (NLRB) case data for the three year periods for which we use CPS data.\footnote{Our thanks to Hank Farber for providing this data.} We focus on certification elections and cases where a conclusive decision on certification was reached.\footnote{As opposed to the case being dismissed or withdrawn.} We use the county of the unit involved in the election to construct our geographic measures, aggregating counties to our city definition discussed above. We also use this data to construct expected probabilities of a firm facing a unionisation election. Unfortunately, the elections data ends before 2020 and so we estimate the full model over the years 1980, 1990, 2000 and 2010.

3.2 Empirical Implementation with Worker Heterogeneity

In order to take the model to the data, we must confront the fact that, while workers are homogeneous in our model, they are not in our data. Our approach is to treat individuals as representing different bundles of efficiency units of work and to assume those bundles are perfect substitutes in production. We then interpret $w_{ict}^n$ in equation (5) as the cost per effective labour unit. Let effective labour units be $\exp(H_l'\beta_t + a_t)$, where $H_l$ and $a_t$ capture observable and unobservable skills of worker $l$, respectively. Adding industry, city and time subscripts, workers’ observed nonunion log wages, $\ln w_{ict}^n$, are given by:

$$\ln w_{ict}^n = H_l'\beta_t + \ln w_{ict}^n + a_t.$$  \hspace{1cm} (17)

The values of $\ln w_{ict}^n$ are our object of interest. To obtain a measure of these, we estimate (17) capturing $\ln w_{ict}^n$ as the coefficients on a complete set of industry×city fixed effects. Our specification of $H_l$ includes a complete interaction of dummies for educational attainment, a quadratic in potential experience, and gender and race dummy variables. We estimate (17) using only nonunion workers, separately by year. This allows for flexible changes over time in the returns to education, etc.. The estimated vector of coefficients on the industry-city fixed-effects are regression-adjusted, average local industry wages, and we use these coefficients as our dependent variable in the regressions (11) and (16).\footnote{As described in section 3.5, we work with an adjusted version of our wage specification in order to move to working with mean log wages.} This procedure removes skill and demographic variation from our wage measure. We also form an analogous regression adjusted union wage measure, $\ln w_{ict}^u$, using only data on unionised workers.

3.3 The Outside Option Terms

Central to our empirical work are the outside option terms characterising alternative job prospects in either the union or nonunion sectors. As defined above, these terms are composed of the rents a worker would get in expectation when searching for a new job ($\sum_\imath u_i w_{ic}^u$)
for union jobs and \( \sum_{i} \eta_{ic} w_{ic}^{n} \) for nonunion jobs) and the probability of finding work in the union sector \( T_{jc}^{u} \), for a worker formerly in a nonunion job in sector \( j \). We use our regression adjusted wages in order get as close as possible to rents rather than skill differentials since wage differences that reflect skill differentials cannot be used as an outside option in bargaining (a janitor cannot use the opening of new jobs for lawyers in town to bargain a better wage).

In our model we assume that the relative likelihood of finding work in the union sector differs by city, previous industry of work, and whether the worker was previously unionised. Using matched CPS data, sample sizes are not sufficient to estimate a fully flexible characterisation of transition paths in this manner. However, working at the national level, we can construct \( T_{it}^{n} \) – the probability that a worker observed in a nonunion job in industry, \( i \), in year, \( t \), who separates from that job in the following year is observed in a union job (in any industry) in year \( t + 1 \). We construct the same probabilities for workers starting in union jobs, creating all of these transition rates separately for each of our sample periods.\(^{24}\) We then combine these rates with the proportion of jobs in the city that are unionised, \( P_{c} \), in order to capture local variability in finding unionised jobs.

More specifically, we construct our transition rate measures as:

\[
T_{ic}^{u} = \frac{T_{i}^{u}P_{c}}{T_{i}^{u}P_{c} + (1 - T_{i}^{u})(1 - P_{c})}, \quad \text{and} \quad T_{ic}^{n} = \frac{T_{i}^{n}P_{c}}{T_{i}^{n}P_{c} + (1 - T_{i}^{n})(1 - P_{c})}. \quad (18)
\]

This is a simplified version of similar measures constructed by \(^{25}\) who uses rich data to calculate transitions between industries. As in \(^{25}\), we interpret this as a measure of relative mobility into the union sector.\(^{25}\) Thus, for a construction worker in Detroit, for example, it measures the relative likelihood of matching with a union job (in any industry) compared to moving into either a union or nonunion job (in any industry). If there are many union jobs in Detroit, and construction workers tend to transition into unionised employment, then these workers will have a higher relative mobility measure of finding unionised employment compared to workers in cities with lower rates of unionisation and who work in industries with relative low transition rates into unionised jobs.

Working with the wage rent variables and the transition rates, we form our measure of the outside option value as:

\[
E_{jc}^{n} = (1 - T_{jc}^{u})\bar{w}_{c}^{n} + T_{jc}^{u}\bar{w}_{c}^{u}, \quad (19)
\]

where \( \bar{w}_{c}^{n} = \sum_{i} \eta_{ic} w_{ic}^{n} \) is the mean, residualized nonunion wage in the city and \( \bar{w}_{c}^{u} \) is the mean, residualized union wage. Thus, the outside option for a nonunion worker in industry \( j \) in city \( c \) is the sum of the probability the person gets a nonunion job times the expected rent from a nonunion job and the probability they get a union job times the expected rent for those jobs.

\(^{24}\)Our framework assumes that bargaining effects operate only through the unemployment channel, that is, workers must first transition through unemployment to access other jobs. However, due to data limitations, our transition measures use transitions between sectors which may, or may not, have included an intervening unemployment spell. Thus, the union outside option term may reflect on-the-job search dynamics. Formally modelling on-the-job search, or job laddering is beyond the scope of this paper. As noted by \(^{25}\) it is not straightforward and is sensitive to modelling of the search process and its relationship to wage determination.

\(^{25}\)Tschopp’s specification would take the same form if the economy was composed of two sectors.
Changes in the outside option are driven by five factors. The first is changes in the local composition of nonunion jobs as captured in the employment shares, i.e., the $\eta_{nic}$'s. In essence, if high-paying jobs such as those in the steel industry are replaced with lower paying service sector jobs then the outside option for all workers in the city is reduced. The second factor is the wage rents in the nonunion sector: even if there is no shift in the industrial composition of nonunion jobs, if the steel industry stops paying higher wages then it no longer offers an attractive outside option for workers in other industries. The third factor is the probability the nonunion worker can get a union job. If union jobs pay, on average, higher rents then a decrease in the probability of getting a union job means lower access to those rents and, therefore, a less valuable outside option. The value of the union option is also altered if there is a shift in the composition of union jobs (the $\eta_{nic}$'s) or the wages in the union sector, which are the fourth and fifth factors.

3.4 Constructing Emulation Probabilities

Since the emulation probabilities are unobserved in our data, we construct a proxy. To do so, we use information from the NLRB and the County Business Patterns (CBP). The idea is to proxy the threat of emulation, $P(\psi_{icf}, \lambda_{ct})$, by the likelihood of that a firm in a given $ic$ cell faces a union election, taking into account the number and composition of establishments at the city-industry level. To this end, our proxy is constructed by first calculating the ratio of the predicted number of elections to the number of establishments in an $ic$ cell. We use this ratio to calculate the probability that an establishment had at least one election in the past four years. The predicted number of elections come from a negative binomial regression, where the observed number of elections are regressed on a number of variables at the $ic$ and city level. These variables include polynomials in the number of workers and the size of establishments in $ic$ along with the unemployment rate, the participation rate, and the average age at the $c$ level plus year- and city-fixed effects. The predicted number of elections come from the fitted values of the model. The details of our procedure are contained in Appendix B.1.

3.5 Dealing with Endogeneity

We will estimate our derived specification in first differences in order to eliminate any industry×city-time invariant characteristics. Ignoring selection issues for the moment, this means in the case without firm responses we consider the regression:

$$\Delta \ln w_{ict}^n = \gamma_{0it}^n + \gamma_1^n \Delta E_{ict}^n + \gamma_2^n \Delta ER_{ct} + \gamma_4^n \Delta \epsilon_{ict}.$$  (20)

where we work with an adjusted version of equation (5) in which we average across firms, add a time subscript, and divide both sides by a base wage, $w_0$, in order to have a dependent variable in log differences.\footnote{As described in section 3.2, we create our dependent variable as residualized mean wages in an $ic$ cell from log wage regressions. We work with log wages in order to match the human capital literature and to make our estimated coefficients easier to interpret.} Given that this specification includes a complete set of industry×time-period fixed effects, the relevant identifying variation for the estimated coefficients comes from across-city within-industry variation. Intuitively, this means that we
identify the impact of outside options by comparing wage changes in the same industry in two different cities that are experiencing different changes in the quality of outside employment prospects, holding the employment rate constant. Given the inherent reflection problem, due to the fact that $E_{ict}^n$ includes local, mean nonunion wages, we would not expect OLS estimation to yield consistent estimates of $\gamma^n_1$. To eliminate this mechanical correlation and to address other potential identification threats, we utilize an instrumental variables strategy as follows.

We begin the construction of our instruments by replacing local wages in $E_{ict}^n$ with what we refer to as industry-level rents. In particular, we estimate separate wage regressions for each of our set of sample years at the national level, working with pooled union and nonunion workers. The regressions include the same set of skill and demographic variables used when forming our residualized wages for the dependent variable plus a complete set of industry dummy variables interacted with a union dummy. We interpret the coefficients on the industry dummies as rents that are allowed to differ in the union and nonunion sectors. Using them, we form:

$$\tilde{E}_{ict}^n = (1 - T_{ict}^n) \sum_i \eta_{ict}^n \nu_{it}^n + T_{ict}^n \sum_i \eta_{ict}^n \nu_{it}^n,$$

where $\nu_{it}^n$ is the wage premium for nonunion workers in industry $i$. Because the premia are at the national level, they break the direct reflection link. The result is an outside option variable expressed such that workers in cities with a concentration in industries that pay high rents at the national level are able to bargain high wages.

In order to use this version of outside option values as the basis for an instrumental variables strategy, we consider changes in each of the right-hand-side terms. The change in the first term equals:

$$\Delta \left( (1 - T_{ict}^n) \sum_i \eta_{ict}^n \nu_{it}^n \right) = \left( (1 - T_{ict}^n) \sum_i \eta_{ict}^n \nu_{it}^n \right) - \left( (1 - T_{ict-1}^n) \sum_i \eta_{ict-1}^n \nu_{it-1}^n \right).$$

Our concern is that movements in the $\eta_{ict}$’s and $T_{ict}^n$ could be correlated with the productivity changes in the error term since sectors with greater increases in productivity will increase their share of employment. Following the standard Bartik approach, we construct predicted values of each that depend on the start of period industrial composition of the city combined with national level changes. That is, we define an instrument:

$$IV_{1ict}^n = \left( (1 - \hat{T}_{ict}^n) \sum_i \hat{\eta}_{ict}^n \nu_{it}^n \right) - \left( (1 - T_{ict-1}^n) \sum_i \eta_{ict-1}^n \nu_{it-1}^n \right),$$

where $\hat{\eta}_{ict}$ corresponds to predicted values of end of period industrial shares in city $c$ for nonunion jobs, and $\hat{T}_{ict}^n$ is the predicted end of period probability of a nonunion worker finding a union job. We define a second instrument, $IV_{2ict}^u$, corresponding to the second term in (21) analogously. We construct predicted employment levels using start-of-period levels

\[27\text{We define the industry dummy variables such that the coefficient values are defined relative to the overall average wage.}\]
at the industry-city level combined with national level growth rates for the relevant industry. Likewise, we combine national level predictions in employment growth with local start-of-period industrial-union composition to form $\hat{T}_{n}^{jct}$, which captures the idea that if there are declines in union employment in sectors with high local employment, then this will predict a decline in the local union proportion. We spell out the details in the construction of $\hat{\eta}_{nict}$ and $\hat{T}_{n}^{jct}$ in Appendix B.3.

To understand the conditions under which $IV_{1}^{n}jct$ and $IV_{1}^{u}jct$ are valid instruments, recall that we include a complete set of time×industry effects and, so, we are working with within-industry, cross-city variation. Note also that the variation in the instrument comes from start of period, cross-city differences in the industrial proportions (the $\eta_{ict}$’s) and the start of period, cross-city differences in the relative shares of union employment in industries (as reflected in $\hat{P}_{ct}$). Validity of the instrument requires that these are uncorrelated with the relevant variation in the error term: cross-city variation in productivity growth. That is, we require an assumptions that the productivity process follows a random walk (since, as BGS show, the $\eta_{ict}$’s can be written as functions of the $\epsilon_{ict}$’s). We can assess this assumption using an over-identification test which we discuss when we present our results.\textsuperscript{28}

For reasons that will become apparent when we discuss addressing selectivity, we are also interested in instruments that do not rely on the predicted change in $\hat{P}_{ct}$. Thus, we form a second set of instruments given by:

$$IV_{2}^{n}jct = \left(1 - T_{jct-1}^{n}\right) \sum_{i} \hat{\eta}_{nict}\nu_{it}^{n} - \left(1 - T_{jct-1}^{n}\right) \sum_{i} \eta_{nict-1}\nu_{it-1}^{n},$$

with an analogous expression for $IV_{2}^{u}jct$. Notice that these instruments rely on the $\hat{\eta}_{nict}$’s but not on $\hat{P}_{ct}$. Because the outside option that a worker uses in bargaining depends on the expected value of wages in other jobs, not on the component parts of that expectation, the two instrument sets should yield the same estimates. That is, it doesn’t matter whether a worker’s outside option worsens because a high rent paying industry leaves town (as emphasized in the difference between $\eta_{jct-1}^{n}$ and $\hat{\eta}_{nct}$) or because of de-unionisation that implies a loss of access to union rents (as captured in the difference between $T_{jct-1}$ and $T_{jct}$). $IV_{2}^{n}jct$ uses only the industry composition change variation while $IV_{2}^{u}jct$ uses both and, theoretically, they should both provide the same answer even though part of the basis of their variation is quite different.

The validity of our instruments relies on the exogeneity of the start of period industry shares and union shares. Our theory implies over-identifying restrictions that allow us to test that exogeneity which we present in the Results section. But we can also follow the advice in ? about checking patterns and correlations for further, suggestive evidence that the exogeneity requirements are met in our case. These checks are weakened to some extent by the fact that our situation differs from the classic Bartik case because our key endogenous variables (the outside option terms) vary at the $ic$ level rather than just the $c$ level. They also include different national level variables in the same expression, requiring us to make

\textsuperscript{28}As ? and BGS point out, Bartik instruments are functions of the start of period values for the $\eta_{ict}$’s – the local industrial composition – and any combination of those values can be used as an instrument. BGS argue that in our case one can find specific combinations within the theory by examining decompositions of the outside option variables that both have intuitive appeal and imply testable over-identifying restrictions.
restrictive assumptions (such as that the industry wage premia are the same in the union and nonunion sectors) in order to fit into the \( ? \) framework. When we do that, the Rotemberg weights for each of our four instruments – weights showing which industries are the main drivers behind the variation in our instruments – point to the top five weighted industries including: mining; motor vehicles and equipment; retail trade; construction; and lumber and wood products. Apart from retail trade, this list is reassuring because it consists of sectors with high wage premia that, at least at one time, were highly unionised. Thus, they seem like a good set of industries for identifying the impacts of variation in access to high rent jobs. \( ? \) also suggest looking for correlates of the baseline industry proportions on which our instruments are built to see if those suggest possible issues. In our case, we would be worried about correlations with variables that might predict growth in city-level productivity. Given that we control for education, age, gender and the employment rate, we do not have candidates for other variables that could fit this bill.

4 Results

4.1 Descriptive Patterns

Before turning to estimation we first present key patterns in unionisation over our sample period. As is well known, the decline of unionisation in the United States over 1980-2019 (and for other rich world nations over a similar time frame, see \( ? \) and \( ? \)), has been remarkable.

In Figure 1 we plot the fraction of workers unionised at the city level over 1980-2019 for each city, highlighting a subset of cities with particularly large or small declines in unionisation. We also highlight the national average (the thick black, dashed line) in the figure. On average, about 25% of jobs were unionised at the city level in 1980, but this number declines to 17% by 1990 and then to 13% by 2019. In cities, like Detroit, Gary, and Pittsburgh, where the union sector played a much larger role in the 1980 economy, the declines are substantial: respectively 21, 29, and 22 percentage points by 2019. Smaller declines (under 10%) are observed in cities with low initial rates of unionisation, such as Dallas, Washington, and Rochester. Thus, there is a considerable range in the changes in unionisation across cities and, importantly, there is variation in the decade in which the declines occur. This will allow us to separate the effects of union declines from general trends.
Observed declines in rates of unionisation are large, which naturally has implications for our outside option terms. In particular, as discussed in Section 3.3, outside options depend on the probability of a nonunion worker getting a union job, $T_{ict}^{u}$, and the average wage in the union sector, $\bar{w}_{ct}^{u}$. Similarly, as union rates fall, so does the threat of unionisation, through $P(\psi_{icf}, \lambda_{ct})$. The theoretical framework we present above suggests that shifts in these terms will impact nonunion wages through both bargaining and threat channels.

As a first step in establishing the relevance of these channels, Figure 2 plots the 2010-1980, city-level change in log nonunion wages against: (a) changes in nonunion-to-union transition rates, $T_{ict}^{u}$; (b) changes in average wages in the union sector, $\bar{w}_{ct}^{u}$; and (c) changes in the probability of a firm facing a union certification election, $P(\psi_{icf}, \lambda_{ct})$. Each of the three panels shows a strong, positive association between changes in the nonunion wage and changes in the outside option and threat components.

In our empirical work, we attempt to separately identify spillover effects from bargaining and threat channels. To do so, we require that cities experience different declines in nonunion-to-union transition rates (that help identify bargaining effects) and union election rates (that help in identifying threat effects). Figure 2 highlights several cities to emphasise this variation in the data. Cleveland and Pittsburgh, for instance, both faced relatively large declines in transition rates (panel A), but very different change in election probabilities.
Likewise, Sacramento experienced large declines in transition rates compared to Dallas, but have similar changes in election probabilities. Overall, the variability across cities in movements in outside option and threat components will help separately identify the bargaining effect from the threat effect.

4.2 Estimation Results

We now turn to estimation results based on the models presented in section 2, beginning with the simple specification (11) and then moving to the specification that includes emulation effects (16). As described in Section 3.2, we work in ten year differences and use the residualized wage as the dependent variable.

4.2.1 Simple Specification

Our simple specification relates changes in nonunion wages to changes in the values of outside options for nonunion workers and the employment rate. It is useful, however, to begin with a more standard specification in order to see how our results relate to the existing literature. In papers including ?, ?, and ?, the main regression takes the form:

\[
\ln w_{icf}^n = a_0 + a_1 P_{ic} + a_2 x_{icf} + u_{icf}
\]  

(22)

where, \(P_{ic}\) is the proportion of workers in the i-c cell who are unionised\(^{30}\), \(x_{icf}\) is a vector of other controls, \(a_2\) is a parameter vector of the same length as \(x_{icf}\), and \(u_{icf}\) is an error term. In earlier papers, the \(x_{icf}\) vector typically includes local demand and supply shifters, such as the proportion of teenagers in the region, the unemployment rate, local lagged per capita income growth, and average firm size, amongst others. Additionally, more recent papers by ? and ? control for full industry-year and/or city-year fixed effects in order to better account for relevant omitted variables.

There are strong similarities between the specification given by equation (22) and our specification (given in equation (11)). In particular, previous studies have included controls similar to the employment rate and the industry-year effects that are in equation (11). The main difference is that union effects are represented by the simple proportion union variable in equation (22) whereas in our specification, theory indicates that they should be part of the outside option term, interacting with changes in union wage premia. The specification does not contain the union proportion on its own.

In the first column of Table 1, we present the results from estimating the standard specification, (22), using our data. Given the way we created the dependent variable, we are controlling for education, age, gender and race in a flexible way. We do not include industry controls, in order to match the older part of the literature. We include changes in the union proportion in an industry-city cell as our core explanatory variable, finding a positive and statistically significant relationship between growth in nonunion wages and growth in unionisation that is broadly similar to previous estimates. ? find a coefficient estimate on the union proportion in the region of .25-.58 for nonunion workers when exploiting industrial

\(^{29}\)Or is pooled for union and nonunion workers, with the proportion union interacted with a union dummy.

\(^{30}\)Typically this metric is calculated either at the national-industry level, or at the city/state level.
Figure 2: Wages and de-unionisation

Notes: Data from the CPS and the NLRB. In all three panels the y-axis denotes the change in regression adjusted nonunion wages at the city level between 2010 and 1980, and the marker size is relative to the size of the city in 1980. In panel A the x-axis variable is the change in our measure of nonunion-to-union transition rates, $T_{ijct}$, aggregated to the city level. In panel B, the x-axis variable is the change in the average, regression adjusted union wage at the city level. In panel C, the x-axis denotes the change in the probability of a firm facing a union certification election calculated from the NLRB data and aggregated to the city level. Appendix B contains more information on our data construction.
variation, although they find little evidence of a spillover effect operating at the local level. find a positive spillover effect using the rate of unionisation at the SMSA for white males, although his results are sensitive to the inclusion of supply and demand shifters and the sample time frame.

Table 1: Non-Union Wages and Outside Options: OLS and 2SLS Estimates

<table>
<thead>
<tr>
<th></th>
<th>OLS</th>
<th>2SLS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1) (2) (3)</td>
<td>(4) (5) (6) (7)</td>
</tr>
<tr>
<td>$\Delta P_{ic}$</td>
<td>0.26***</td>
<td>0.014</td>
</tr>
<tr>
<td></td>
<td>(0.024)</td>
<td>(0.015)</td>
</tr>
<tr>
<td>$\Delta((1 - T_{ict}^N) \cdot \bar{w}_{ct}^N)$</td>
<td>0.92***</td>
<td>0.69***</td>
</tr>
<tr>
<td></td>
<td>(0.016)</td>
<td>(0.074)</td>
</tr>
<tr>
<td>$\Delta(T_{ict}^N \cdot \bar{w}_{ct}^U)$</td>
<td>0.80***</td>
<td>0.64***</td>
</tr>
<tr>
<td></td>
<td>(0.029)</td>
<td>(0.063)</td>
</tr>
<tr>
<td>$\Delta E_{ict}^N$</td>
<td></td>
<td>0.65***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.067)</td>
</tr>
<tr>
<td>$\Delta E_{R}$</td>
<td>-0.091*</td>
<td>0.14</td>
</tr>
<tr>
<td></td>
<td>(0.051)</td>
<td>(0.10)</td>
</tr>
<tr>
<td>Observations</td>
<td>9024</td>
<td>9024</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.026</td>
<td>0.026</td>
</tr>
<tr>
<td>Year × Ind.</td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
<td>Instrument set:</td>
<td>IV$^{1N}_{jct}$</td>
<td>IV$^{1N}_{jct}$</td>
</tr>
<tr>
<td>First-Stage p-Stat.:</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$\Delta((1 - T_{ict}^N) \cdot \bar{w}_{ct}^N)$</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td>$\Delta(T_{ict}^N \cdot \bar{w}_{ct}^U)$</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td>$\Delta E_{ict}^N$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Over-id. p-val</td>
<td>.</td>
<td>.</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0.457</td>
</tr>
</tbody>
</table>

Notes: This table displays results from the estimation of equations (22) (columns 1 and 2) and (5) (columns 3 - 7) via OLS (columns 1 - 3) and 2SLS (columns 4 - 7). The dependent variable is the decadal change in the regression adjusted average hourly wage of nonunion workers in an industry-city cell, using CPS data from 1980-2019 across 50 industries and 93 cities. Standard errors, in parentheses, are clustered at the city-year level.

In column (2), we control for industry-year fixed effects which leads to a large decrease in the magnitude of the coefficient on the proportion union from .26 to .03. In comparison, employ separate industry and year effects, working with national level variation. They estimate a negative effect of the union proportion on nonunion wages. exploits variation in the probability of unionisation across states and industries in the cross-section, and in turn controls for state and industry fixed effects. He finds a coefficient estimate on the probability of unionisation around .18 which declines significantly over time. This result is
more similar to ours than that of ?, but in both cases the source of variation is not directly comparable. In taking first differences we are controlling for any fixed city-industry effects, and we additionally control for common industry trends in wage growth.

In the third column of Table 1, we present OLS estimates of our simple specification but (contrary to that derivation) include the proportion union in the \(ic\) cell. We break the outside option term into its two components: the probability of finding a nonunion job times the expected wage in nonunion jobs \((1 - T^n_{jc}) \sum_i \eta^n_{ic} w^n_{ic}\); and the probability of finding a union job times the expected wage in those jobs \((T^n_{jc} \sum_i \eta^n_{ic} w^u_{ic})\). Note that the coefficients on these variables should be equal since what matters is the overall outside option value and these are just components of that value. In other words, it should not matter to a firm whether a worker’s outside option loses value because a high rent unionised firm leaves town or a nonunionised firm in an industry that also pays high rents shuts down.

The results in column (3) indicate that the option values associated with union and nonunion jobs have positive and significant effects on nonunion wages and are similar, though certainly not identical, in size. In addition, the employment rate enters significantly but with the theoretically incorrect sign. In column (4), we present results of IV estimation using the \(IV^1_u\) and \(IV^1_u\) instruments. We do not instrument for \(\Delta ER_{ct}\) even though there are clear reasons to assume it is correlated with our error term.\(^{31}\) We follow ? in interpreting the employment rate as a control variable - a variable that is not of direct interest in its own right but is useful for picking up its own effect and those of correlated omitted variables. In our case, we view the employment rate as capturing its own effect plus the impact of general, local demand shifts. This allows us to isolate the outside option effects we care about from demand effects. We present the \(p\)-values for the Sanderson-Windmeijer test statistics for weak instruments (?) at the bottom of the table. These take values of 0.001 or less in all cases, indicating that we do not face weak instrument problems.

The results in column (4) again show positive and significant outside option effects, with the estimated coefficients for the two components being very similar in size. This and the fact that the employment rate now takes the correct sign are strongly supportive of the model since there is no mechanical reason why the two outside option terms should have similar sized effects. In the context of the model, in which the composition of local employment does not determine productivity within a specific industry, the significance of these effects implies that wages are partly driven by bargaining responses to rents in the local economy. Following BGS, we view the fact that we get these estimates while controlling for industry specific trends and controlling for the local employment rate as reinforcing this interpretation. Note, further, that the coefficient on the change in the union proportion is now 0.014 and smaller than its standard error. This suggests that any threat effects captured by the union proportion variable occurs through the bargaining channel reflected in our outside options variables. In column (5), we present our specification, dropping the proportion union, with no impact on our key estimated coefficients.

\(^{31}\)We did implement a specification in which we instrumented for \(\Delta ER_{ct}\) using the classic Bartik, labour demand instrument. We find that this instrument is highly correlated with our other instruments and results in large estimates of the impact of the employment rate that are out of line with those in BGS. The estimates for the proportion union and the outside option value coefficients are very similar regardless of whether we instrument for the employment rate. The results in which we instrument for the employment rate can be found in Appendix C.
To understand the magnitude of our estimated outside option coefficients, recall from equation (11) that $\gamma^n_i$ is the impact of a 1 dollar increase in the expected value of the outside option for a nonunion worker in industry $i$ in city $c$ on that worker’s wage. But this is only the immediate impact of a shift in, say, the unionisation rate. The resulting increase in the wage in $i$ is an increase in the outside option for workers in other industries, inducing further increases in their wages that then imply an increase in the value of the outside option and the wage in industry $i$, which implies a further increase in the outside option for other industries, and so on. In the end, the total impact of a 1 dollar increase in the outside option value for the mean wage in industry $i$ is $\frac{\gamma^n_i}{1-\gamma^n_i}.32$ Thus, our estimated initial impact of a one unit change in the value of the outside option of 0.66 (the average of the two estimates in column (5)) becomes 1.78 once we include feedback loops of the spillovers.

Following BGS and ?, we can discuss this total effect in relation to a standard shift share estimate of the impact of a change in the industrial composition of nonunion jobs in a city on the mean nonunion wage, $\bar{w}_{ct}^n$. In particular, we can decompose the change in $\bar{w}_{ct}^n$ as:

$$\Delta \bar{w}_{ct}^n = \sum_i \Delta \eta_{ict}^n w_{ict}^n + \sum_i n_{ict+1} \Delta w_{ict}^n$$

(23)

The first term on the right hand side is the ‘between’ component, showing the effect of changes in the industrial composition. The second term is the ‘within’ component, showing the effect of changes in wages within sectors. For a clean decomposition, these two terms are assumed to be independent. But in our model, a change in composition has both its direct ‘between’ effect and an effect through inducing changes in wages within industries because of the alteration in the outside option value. Thus, if the industrial composition changes by enough to increase the ‘between’ component by 1 unit, in a standard setting, the average nonunion wage in the city will also increase by 1 unit. With bargaining induced spillovers, we also need to add in the effects on wages within industries, noting from (19) that a 1 unit increase in the mean nonunion wage would increase the outside option value, $E_{jc}$ by $(1 - T_{jc})$ units. Thus, the total effect on wages would be $1 + \frac{\gamma^n_i}{1-\gamma^n_i} \cdot (1 - T_{jc})$ units. Using the average of the two outside option coefficients in column (5) and an average value of $T_{jc}$ of 0.04, this implies that to get the total impact of a change in industrial composition on the mean nonunion wage we would multiply the standard ‘between’ component measure of that impact by 2.86. This is very similar in size to what BGS found not taking account of the union sector.

In column (6), we estimate a specification in which use the IV2 instruments. Recall that these instruments do not use variation in the local proportion of workers who are unionised. The fact that we get very similar results with these instruments and the IV1 instruments means that changes in outside options induced by either changes in the probability of accessing union jobs or the industrial composition of jobs have the same effect. This is implied by the theory (since firms don’t care why the outside option value changes) and serves as an over-identifying test of the model. It also implies that, when it comes to bargaining, de-unionisation is not special – it is a reason that worker bargaining power is reduced just like the decline in any other high wage option they might lose. As we will discuss in the next

\[32\] Here, we have assumed that $\gamma^n_u = \gamma^n_1$ so that the spillover effects on union and nonunion wages are the same.
section, the fact that we get the same results when we don’t use variation in the proportion unionised is also useful when we investigate potential selection issues.

In column (7), we impose the restriction that the coefficients on the two components of the outside option take the same value in an estimation using the IV2 instruments. Not surprisingly, given the similarity of the coefficients in column (6), we cannot reject the restriction at any standard level of significance (the $p$-value associated with testing the restriction is given at the bottom of column (5)). We view this as another over-identifying test of our model since, again, there is no mechanical reason why the two terms should have the same effect (and they are based on quite different variation – in nonunion versus union industrial proportions and wage premia) but theoretically their effects should be identical.

4.2.2 Controlling for Selectivity

Thus far we have presented estimates using IVs to break the linkage between local productivity shocks and growth in outside options. Section 2.5, however, makes clear that there is likely selectivity into the union sector based on productivity draws. This arises because productivity shocks are differently weighted in the union and nonunion sector due to alternative methods of wage bargaining.

We address selection through a generalized Heckman two-step approach (see ?, ?, ?). The idea in this approach is that the error mean term in (11), $E(\epsilon_{ic}|I_{icf} \leq 0)$, creates an omitted variables bias that can be addressed by including a fitted error mean term as a covariate in the regression. Further, the error mean can be expressed as a non-linear function of the probability of selection (the probability of being nonunion in our case) or of exogenous variables that drive that probability. The fact that the error mean term is a function of $P_{ict}$ implies that instrumental variables approaches to estimating the standard specification are invalid in the presence of selection. In that situation the variation in $P_{ict}$ induced by an instrument identifies a coefficient that is a combination of the causal effect of the union proportion on wages and the fact that when $P_{ict}$ changes, the composition of nonunion firms (and the mean of the error term) necessarily changes. However, our instrument set includes instruments that are not a function of the union proportion and we have seen that we obtain very similar results whether or not we include the union proportion and union transition rate variables as instruments. Thus, we can take an approach in which we use the restricted instrument set that does not include the union proportion to identify the effects of our rent variables, using functions of the union proportion or related variables to absorb the selectivity effect.

Given these arguments, we examine potential selection effects using two sets of variables. First, we include a quartic in the change in the proportion of workers in the industry×city×time cell who are unionised. In doing this, we are taking the model very seriously. In particular, we are taking advantage of the difference between the transition probability of workers (used in the construction of the outside options), and the proportion of workers who are unionised at a point in time. The latter is the theoretically correct variable for dealing with selection since differences in the proportion who are unionised is directly related to differences in selectivity. We present both OLS and IV estimates of our main specification including the quartic in the change in the union proportion. For the IV estimates, we use the IV2 set – the instruments that do not include any changes in the unionisation rate.
Table 2: Non-Union Wages and Outside Options: Controlling for Selectivity

<table>
<thead>
<tr>
<th></th>
<th>OLS</th>
<th>2SLS</th>
<th>OLS</th>
<th>2SLS</th>
</tr>
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<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
</tr>
<tr>
<td>( \Delta((1 - T_{ict}^N) \cdot \bar{w}_{ct}^N) )</td>
<td>0.92***</td>
<td>0.61***</td>
<td>0.92***</td>
<td>0.61***</td>
</tr>
<tr>
<td></td>
<td>(0.016)</td>
<td>(0.100)</td>
<td>(0.020)</td>
<td>(0.11)</td>
</tr>
<tr>
<td>( \Delta(T_{ict}^N \cdot \bar{w}_{ct}^U) )</td>
<td>0.80***</td>
<td>0.69***</td>
<td>0.82***</td>
<td>0.72***</td>
</tr>
<tr>
<td></td>
<td>(0.030)</td>
<td>(0.089)</td>
<td>(0.032)</td>
<td>(0.080)</td>
</tr>
<tr>
<td>( \Delta E_{ict}^N )</td>
<td></td>
<td>0.65***</td>
<td></td>
<td>0.68***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.067)</td>
<td></td>
<td>(0.065)</td>
</tr>
<tr>
<td>( \Delta E_{ict}^R )</td>
<td>-0.090*</td>
<td>0.23*</td>
<td>0.19*</td>
<td>-0.028</td>
</tr>
<tr>
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<td>(0.13)</td>
<td>(0.097)</td>
<td>(0.072)</td>
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<td>9024</td>
<td>9024</td>
<td>6860</td>
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<td>0.21</td>
<td>0.22</td>
<td>0.54</td>
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<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
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<td>( IV_{jct}^{2N} )</td>
<td>( IV_{jct}^{2N} )</td>
<td>( IV_{jct}^{2N} )</td>
<td>( IV_{jct}^{2N} )</td>
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<tr>
<td></td>
<td>( IV_{jct}^{2U} )</td>
<td>( IV_{jct}^{2U} )</td>
<td>( IV_{jct}^{2U} )</td>
<td>( IV_{jct}^{2U} )</td>
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<tr>
<td>First-Stage F-Stat.:</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>( \Delta((1 - T_{ict}^N) \cdot \bar{w}_{ct}^N) )</td>
<td>0.000</td>
<td>0.000</td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \Delta(T_{ict}^N \cdot \bar{w}_{ct}^U) )</td>
<td>0.000</td>
<td>0.000</td>
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<tr>
<td>( \Delta E_{ict}^N )</td>
<td></td>
<td>0.000</td>
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<td>0.000</td>
</tr>
<tr>
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<td>0.474</td>
<td>.</td>
<td>0.326</td>
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</table>

Selection Controls

- \( P_{ic} \) Quartic: Yes Yes Yes No No No
- Election Vars: No No No Yes Yes Yes

Joint Tests:

- \( p \)-val: .855 .867 .8214 .17 .306 .1732
- \( F \)-Stat: .334 .316 .3821 1.54 1.2008 1.5148

Notes: This table displays results from the estimation of equation (5) via OLS (columns 1 and 4) and 2SLS (columns 2,3,5, and 6). The dependent variable is the decadal change in the regression adjusted average hourly wage of nonunion workers in an industry-city cell, using CPS data from 1980-2019 across 50 industries and 93 cities. Standard errors, in parentheses, are clustered at the city-year level.
Following ?, we also estimate a specification in which we proxy for costs of unionisation using National Labor Relations Board (NLRB) data on certification elections. In particular, we calculate the number of workers involved in certification elections in each city over three year windows around the years 1980-1990-2000-2010, divided by the number of nonunion workers in the city. We also construct a second measure: the fraction of certification elections won over the same three year window. We see low values of both variables as reflecting higher costs to successful unionisation in a location. We use this to address selection by including quadratics in both variables and their interaction.

We present results from the specifications using the quartic in unionisation proportions to control for selection in columns (1)-(3) of Table 2, with column (1) containing OLS estimates and the IV estimates in columns (2) and (3). A test of the hypothesis that the parameters in the quartic equal zero is not rejected at any standard significance level, and the estimates for the key covariates change very little from Table 1. Columns (4)-(6) contain the results when we use the NLRB variables to address selection. Because the elections data series does not extend to 2020, we estimate this specification using the 1980 through 2010 data. Here, we can again not reject the null hypothesis of no selectivity effects, though the associated p-values are much lower than with the union proportion quartic. The main estimated effects are more erratic, though the outside option effect is very similar to the estimates in Table 1 if we impose the restriction that the union and nonunion outside options have the same effect. We conclude that selectivity is not a central issue driving our results.

4.2.3 Results Allowing for Firm Response

We now turn to our full specification, (16), in which we allow for nonunion firms to respond to the threat of unionisation by increasing their wages. As we saw in section 2, we do this by multiplying the outside option variable for a nonunion worker, $E_n$, by the probability the firm is not a union emulator and introducing a new variable equal to the outside option value for a union worker, $E_u$, multiplied by the probability the firm is an emulator. We construct the outside option variables as before, with $E_n$ and $E_u$ differing in the transition probabilities to entering union jobs for union and nonunion workers (with the estimated probabilities from the national data being substantially larger for the union workers). The probability of a nonunion firm being a union emulator is not directly observed and we proxy for it using our constructed probability that firms in the $ic$ cell faced a unionisation drive in the previous four years, described in section 3.4. As we saw earlier, the probability of a union election moves differently from the probability a worker can transfer to a union job. We view variation in the probability of an election times the outside option value for union workers as reflecting the traditional threat effect and variation in the probability of a nonunion worker obtaining a union job times the expected rents in those jobs as the bargaining effect. These correspond to the two main right hand side variables in our full specification.

We present the results for the full, firm response model in Table 3. The first column contains IV estimates, using the $IV1$ instrument set along with $\hat{P}(\psi_{icf}, \lambda_c)$ and interactions between it and the $IV1$ instruments as instruments (calling this instrument set $IV3$). Recall from equation (16) that the coefficient on the first term (the probability of not being a union emulator times the expected value of wages for a nonunion worker) is the same as on the outside option value in the simple specification. And, in fact, the estimated coefficient for
that first term is very similar to what we observed in Table 1. For the second term (the union worker outside option value times the probability that the firm is an emulator), the coefficient is the effect of an increase in the outside option value on a union worker’s wage and turns out to be very similar in size to the effect for nonunion workers. In column (2), we replace the IV1 instruments with IV2 instruments and find even greater similarity in the outside option effects for union and nonunion workers at similar size effects to column (1)\textsuperscript{33}. In column (3), we include the quartic in changes in the proportion union to address selection. As in Table 2, the main estimated effects change very little when we control for selectivity.

The key result from these estimates is that both the bargaining channel (as captured in the coefficient in the first row) and the standard threat channel (represented by the coefficient in the second row) matter. In fact, they have similar sized coefficients, though there is no theoretical reason why they would be equal. If $\gamma_1^n = \gamma_1^u$ (as seems to be a reasonable assumption given the estimates) and $T_{ic}^n = T_{ic}^u$ then the impact of a one dollar increase in the mean nonunion wage due to a shift in industrial composition would have the same total effect as in the simple regression (i.e., $1 + \frac{\gamma_1^n}{1-\gamma_1^n} \cdot (1 - T_{jc}^n)$ or about 2.9 times the standard ‘between’ effect computation of the industrial composition shift).\textsuperscript{34} This occurs because the mean nonunion wage is part of both $E_{ic}^n$ and $E_{ic}^u$ and the former is multiplied by $(1 - P(\psi_{icf}, \lambda_c))$ while the latter is multiplied by $P(\psi_{icf}, \lambda_c)$. The significance of the union worker outside option is, in some ways, remarkable. Its effect is identified relative to the nonunion worker outside option only because the transition probabilities to union jobs are larger for union workers.

In column (4), we interact the outside option variables with an indicator variable for whether the city is located in a state with Right to Work (RTW) laws at the start of the decade. Given this specification and the fact that only 6 states adopt RTW laws in our time period (\textsuperscript{?}), this indicator is not about the effects of becoming a RTW state but about whether nonunion wage setting is different in states with ongoing RTW environments. The impact of the nonunion option value is slightly larger in RTW states. But more importantly, adding together the union outside option coefficient with the coefficient on its interaction with the RTW indicator yields a statistically insignificant effect of only 0.24 in RTW states. That is, we cannot reject the hypothesis that there is no traditional threat effect in states where RTW laws eliminate the threat of being unionised. This is both interesting in its own right and a further piece of evidence for our claim that we are identifying emulation effects.

Taken together, our results indicate that de-unionisation affected nonunion wage setting through both bargaining and standard threat channels, though the latter does not operate in RTW environments. The effects appear to be sizeable relative to what one would compute with simple shift-share calculations but we will confirm their overall impact in a counterfactual exercise after investigating heterogeneity across different demographic and skill groups.

\textsuperscript{33}We refer to the set of instruments composed of IV2 and interactions with $P(\psi_{icf}, \lambda_c)$ as IV4.

\textsuperscript{34}If, as we observe, $T_{ic}^u > T_{ic}^n$ then the impact of a one dollar change in the mean non-union wage is smaller by $\gamma_1^u P(\psi_{icf}, \lambda_c)(T_{ic}^u - T_{ic}^n)$ than it would be if $T_{ic}^u = T_{ic}^n$, which is a small adjustment.
Table 3: Non-Union Wages and Outside Options: Including Wage Emulation Effects

<table>
<thead>
<tr>
<th></th>
<th>2SLS</th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
</tr>
<tr>
<td>(\Delta \left( (1 - P(\psi_{icf}, \lambda_c)) \cdot E_{ic}^N \right) )</td>
<td>0.70***</td>
<td>0.69***</td>
<td>0.70***</td>
<td>0.70***</td>
<td>0.68***</td>
</tr>
<tr>
<td></td>
<td>(0.071)</td>
<td>(0.072)</td>
<td>(0.073)</td>
<td>(0.071)</td>
<td>(0.075)</td>
</tr>
<tr>
<td>(\Delta \left( P(\psi_{icf}, \lambda_c) \cdot E_{ic}^U \right) )</td>
<td>0.64***</td>
<td>0.70***</td>
<td>0.63***</td>
<td>0.54***</td>
<td>0.52</td>
</tr>
<tr>
<td></td>
<td>(0.094)</td>
<td>(0.088)</td>
<td>(0.093)</td>
<td>(0.13)</td>
<td>(0.88)</td>
</tr>
<tr>
<td>(\Delta \left( (1 - P(\psi_{icf}, \lambda_c)) \cdot E_{ic}^N \right) \cdot RTW )</td>
<td>0.52</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.035)</td>
<td></td>
<td></td>
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<td></td>
</tr>
<tr>
<td>(\Delta \left( P(\psi_{icf}, \lambda_c) \cdot E_{ic}^U \right) \cdot RTW )</td>
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<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td></td>
<td>(0.18)</td>
<td></td>
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</tr>
<tr>
<td>(\Delta ER_c )</td>
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<td>0.36**</td>
<td>0.34**</td>
<td>0.25*</td>
<td>0.46***</td>
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<tr>
<td></td>
<td>(0.15)</td>
<td>(0.15)</td>
<td>(0.15)</td>
<td>(0.15)</td>
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</tr>
<tr>
<td>Observations</td>
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<td>6860</td>
<td>6860</td>
<td>6860</td>
<td>6860</td>
</tr>
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<td>(R^2 )</td>
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<td>0.254</td>
<td>0.256</td>
<td>0.256</td>
<td>0.251</td>
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<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
<td>Yes</td>
</tr>
<tr>
<td>(P(\psi_{icf}, \lambda_c) \times \text{Ind.} )</td>
<td>Yes</td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Instrument set:</td>
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<td>(IV4^N_{jct}, IV5^N_{jct} )</td>
<td>(IV3^N_{jct} )</td>
<td>(IV3^N_{jct}, IV3^N_{jct} \times RTW )</td>
<td>(IV3^N_{jct} )</td>
</tr>
<tr>
<td></td>
<td>(IV3^U_{jct} )</td>
<td>(IV4^U_{jct}, IV5^U_{jct} )</td>
<td>(IV3^U_{jct} )</td>
<td>(IV3^U_{jct}, IV3^U_{jct} \times RTW )</td>
<td>(IV3^U_{jct} )</td>
</tr>
<tr>
<td>Select controls</td>
<td>(\Delta P_{ic} ) Quartic</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>First-Stage (p)-Stat.:</td>
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<td></td>
<td></td>
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<td></td>
</tr>
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<td>(\Delta \left( (1 - P(\psi_{icf}, \lambda_c)) \cdot E_{ic}^N \right) )</td>
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<td>0.000</td>
<td>0.000</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td>(\Delta \left( P(\psi_{icf}, \lambda_c) \cdot E_{ic}^U \right) )</td>
<td>0.000</td>
<td>0.000</td>
<td>0.000</td>
<td>0.000</td>
<td>0.000</td>
</tr>
<tr>
<td>(\Delta \left( (1 - P(\psi_{icf}, \lambda_c)) \cdot E_{ic}^N \right) \cdot RTW )</td>
<td>0.000</td>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(\Delta \left( P(\psi_{icf}, \lambda_c) \cdot E_{ic}^U \right) \cdot RTW )</td>
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<td>Notes: This table displays results from the estimation of equation (16) via 2SLS. The dependent variable is the decadal change in the regression adjusted average hourly wage of nonunion workers in an industry-city cell, using CPS data from 1980-2019 across 50 industries and 93 cities. Standard errors, in parentheses, are clustered at the city-year level.</td>
<td></td>
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4.2.4 Heterogeneity in Spillover Effects

There is considerable heterogeneity in the unionisation experience. Other evidence indicates that unionisation was particularly prevalent among low-skilled men (although as highlighted by ?, there has been a remarkable rise in the share of unionised jobs held by women). The decline in unionisation might then be expected to have had a bigger impact on these groups, and so it would be useful to know if they have a bigger or smaller reaction to outside options in their wage setting.

In Table 4, we present estimates of the bargaining and standard threat effect coefficients ($\gamma_1^u$ and $\gamma_1^a$ in equation (16)) for a set of sub-populations defined by gender, age, and education. Each row corresponds to estimates for a different subsample. To construct industry by location cells of sufficient size, we shift to using states rather than cities as our geographic dimension. Estimating our specification with the full sample with states generates very similar effects (though slightly larger in absolute value) to those presented in the tables up to this point, which use cities. The 3rd and 4th columns of the table also show the $p$-values from SW weak instrument tests for the instruments corresponding to the two outside option terms. In all cases, these $p$-values are 0.02 or less, implying no weak instrument problems. The sample size varies across the sub-populations because we eliminate ic cells with fewer than 20 observations in each case.

The first two rows contain separate results for men and women. These indicate that both the bargaining and standard threat effects are larger for women, implying that declines in unionisation would have a more negative effect on nonunion wages for women than men. On the other hand, differences by age group are not generally substantial. The other noticeable pattern is the relative importance of the traditional threat effect compared to the bargaining effect for the less educated – and, particularly, for less educated males. While all the effects are statistically significant and of a similar order of magnitude to our whole sample estimates, the traditional threat effect estimate is twice as large as the bargaining effect for the low (high school graduates or less) educated while the opposite is true for those with a post-secondary education. That is the relatively more salient threat for the lower educated is to unionise the current workplace rather than to leave and find a better job, which fits with characterizations of unionisation among less-educated workers as raising wages through organizing relatively homogeneous workers (in terms of skill) who do not have much individual power because their specific outside options are not strong.

We have also examined potential heterogeneity between public and private sector unions. Thus far we have included the public sector, both in the construction of our outside option terms, and as an observation on the left hand side of the wage equation. However outline the marked difference in unionisation between the private and public sectors since 1980 such that unionisation is now 5 times higher in the public sector. To the extent that wage setting is different in the public sector, these shifts in composition could be driving some of our results. In Appendix C, we present results excluding the public sector both in the construction of the dependent variable and in the construction of our outside option variables and associated instruments. Our results are robust to these changes, though our estimated spillover effects are slightly larger in the simple specification.
Table 4: Subsample Analysis - Coefficient Estimates on Outside Options

<table>
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<tr>
<th>Sample</th>
<th>$\gamma_1^n$</th>
<th>$\gamma_2^n$</th>
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<th>(2)</th>
<th>(3)</th>
<th>(4)</th>
<th>(5)</th>
</tr>
</thead>
<tbody>
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<td>Men</td>
<td>0.55**</td>
<td>0.44**</td>
<td>0.00</td>
<td>0.00</td>
<td></td>
<td></td>
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<td></td>
<td>(0.13)</td>
<td>(0.18)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Women</td>
<td>0.73**</td>
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<td></td>
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<tr>
<td></td>
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<td>(0.17)</td>
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</tr>
<tr>
<td>Age 20–26</td>
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<td>0.40**</td>
<td>0.00</td>
<td>0.00</td>
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<tr>
<td></td>
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<td>(0.19)</td>
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<tr>
<td>Age 36–55</td>
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<td>0.67**</td>
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<td>0.00</td>
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<tr>
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<td>≤ HS</td>
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<tr>
<td></td>
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<tr>
<td>&gt; HS</td>
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<td>0.41*</td>
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<tr>
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<td>(0.23)</td>
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</tr>
<tr>
<td>Men: Age 20–26</td>
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<td>0.74**</td>
<td>0.00</td>
<td>0.00</td>
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<td>(0.25)</td>
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<td>(0.68)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Men: ≤ HS</td>
<td>0.52**</td>
<td>1.10*</td>
<td>0.00</td>
<td>0.00</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.14)</td>
<td>(0.62)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Men: &gt; HS</td>
<td>0.85**</td>
<td>0.55*</td>
<td>0.00</td>
<td>0.00</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.16)</td>
<td>(0.33)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Notes: This table displays results from the estimation of equation (16) via 2SLS using 2SLS on separate subsamples.
5 Counterfactual Exercise

Our results thus far indicate a significant relationship between quality job opportunities in both the nonunion and union sectors and nonunion wage setting. However, the exact magnitude of the estimated effects remain unclear. In this section, we pursue a counterfactual exercise, asking what path mean wages in a typical city would have followed if unionisation rates and union wage premia had remained at their 1980 levels. This both provides a way of characterizing the size of our estimated effects and some insight into whether deunionisation played an important role in wage changes over the last four decades.

5.1 Loss of Union Power and Movements in the Average Wage

Our focus is on changes in total mean wages at the city level, expressed as the weighted average of nonunion and union mean wages, where the weight is the proportion unionised at the city level, $P_{ct}$:

$$w_{ct} = P_{ct} \times w_{u ct} + (1 - P_{ct}) \times w_{n ct}$$  \hspace{1cm} (24)

In order to focus attention on trends related to unions, undistracted by changes in education, age, etc., we use the residualized industry-city wages that formed the dependent variable in our regressions combined with industrial shares at the local level to create city level wages (i.e., $w_{u ct} = \sum_i \eta_{u c} w_{ict}$ and $w_{n ct} = \sum_i \eta_{n c} w_{nict}$).

Changes in the strength of unions affect movements in the overall average city wage through four channels. The first is changes in the union proportions, $P_{ct}$. This is the most direct effect of deunionisation: the shift in workers out of higher paid (union) jobs to lower paid (nonunion) jobs, holding constant the wages in each sector. This is the ‘between’ component in standard decompositions and is often presented as the effect of unions on the wage structure. The second channel is the effect of de-unionisation on nonunion wages through the classic threat of unionisation route. In our full wage specification, (16), this is captured by changes in the probability that nonunion firms face union election drives ($P(\psi_{icf}, \lambda_{c})$). The loss of union power also affects nonunion wages through reducing outside options (reductions in bargaining power). This happens, in part, because of reductions in the probability a nonunion worker can find a union job. In our specification, this shows up in two places. The first is a decline in $T^n_{jc}$, which refers to a fall in the probability of getting a union job in any sector. But the impact on worker outside options will obviously be greater if it is mostly union jobs in, say, the high-paying manufacturing industry that are lost. At the same time, we don’t want to assign all industrial changes as union effects. Instead, we assume that shifts in the industrial distribution for nonunion workers (the $\eta^n_{ic}$’s) capture changes in the overall economy while a change in the industrial distribution for union workers relative to what happens for nonunion workers ($\eta^u_{ic} - \eta^n_{ic}$) is a union decline effect. These shifts in $T^n_{jc}$ and the ($\eta^u_{ic} - \eta^n_{ic}$)’s form the third union impact channel.

The value of the outside option for nonunion workers is also affected by declines in the union wage premia (the fourth channel). If unions become less effective at unifying worker resistance during bargaining or afraid to threaten the withdrawal of the whole workforce in a new policy environment then the union wage premium will decline. In that case, the value of the outside option of finding a union job for a nonunion worker also declines. Here, too,
we focus on the relative decline of union versus nonunion wages (the union wage premium).\footnote{A fifth channel is through selection, as shifting firms from being union to nonunion changes the productivity composition of nonunion firms. In our model, this effect implies an increase in the observed nonunion wage, offsetting the negative effects of declining union power operating through the first two channels. However, we find no evidence that selection had substantial effects on observed mean wages and, so, do not include this channel in our decompositions.} Versions of the third and fourth channels also exist for union wage determination: as unions are weakened other union workers face declines in their own outside options.

In Figure 3, we plot percentage changes in these key drivers \( P_c, P(\psi_{icf}, \lambda_c), T^n_{jc}, \) and \( (w^u - w^n) \) relative to their 1980 values. For each of the probabilities, we work with values at the industry-city cell level, aggregating them to a single value using the proportion of all nonunion workers in each cell in 1980. The union wage premium comes from our estimates at the nation level. Our model indicates that the different types of unionisation probabilities are relevant for different parts of the wage setting process. \( P_c \) does not appear directly in our model but is part of the first step of our decomposition, as we discussed earlier, and is what is used in other papers. \( T^n_{jc} \) is relevant for outside options since it shows whether workers are actually able to move into union jobs. Its movements are obviously related to the decline in the proportion of workers who are unionised, though one could imagine it declining either faster than that proportion (if older union workers keep their jobs but new job searchers have difficulty getting into a union job) or slower than that proportion (if the proportion declines quickly because union workers suddenly start taking early retirement). In fact, the figure shows that the two proportions move similarly in the 1980s but the probability of entering a union job declines faster after 1990. \( P(\psi_{icf}, \lambda_c) \) is relevant for the traditional threat effect since it shows changes in whether a firm can expect to face a unionisation drive. It falls the fastest of any of the unionisation measures, particularly in the 1980s when the policy environment was shifting strongly against unionisation. It is worth noting, though, that the probability of a firm facing a union election was small even in 1980 (on the order of 4%).

Perhaps the most interesting line in Figure 3 is the one corresponding to the union wage premium. The premium actually increases in the 1980s before showing a sizeable decline in the 1990s and a smaller one thereafter.\footnote{\textsuperscript{36} They plot union wage premiums over an extended time period. Their plot differs from ours in showing a flat premium over the 1980s but is similar in showing a decline after 1990. Their estimates are based on family income and do not include controls for education that are part of our estimation.} Our model provides an explanation for the increase in the 1980s based on the emulation channel that echoes a discussion in ?. Recall that the observed mean non-union wage equals a weighted average of the pure nonunion wage \( (w^n) \), paid by firms under no threat of unionisation, and the emulation wage \( (w^u^*) \), paid by firms seeking to stave off unionisation. The weights are \( P(\psi_{icf}, \lambda_c) \) and \( (1 - P(\psi_{icf}, \lambda_c)) \), respectively. Suppose that larger forces (trade, technological change, etc) drive down both \( w^n \) and the union wage, \( w^u \), to the same extent. If the threat of unionisation declines at the same time then the observed nonunion wage will fall farther because there will be fewer emulating firms and the emulation wage they have to pay won’t be as high. This pattern of faster decline in mean observed wages in the nonunion sector is what we observe in the 1980s. It is striking that this is the decade in which the union threat (the probability of facing a union election) fell fastest relative to other unionisation probabilities.
Notes: Data from the CPS and the NLRB. Each series is represented as a percentage change from the corresponding 1980 level. Proportion union, union premium and transitions are constructed from the CBS data, discussed in Appendix B. The treat of union election comes from NLRB data, and is described in detail in Appendix B.1.
5.2 Overall Decomposition

We present our decomposition of the overall trend in average city wages in Figure 4. Recall that we are working with residualized wages after taking out education, race, age, and gender effects. We set the wage level to correspond to the mean wage across all worker types. To estimate time trends we average across cities each period, using city populations as weights. Thus, we are showing the components of mean wage movements for an average city, not the components of the trend in mean wages at the national level. The bottom line in Figure 4 is the actual trend in the (residualized) mean wage for an average city. It depicts an overall real wage trend that is strongly decreasing between 1980 and 1990 – falling by approximately 15% in that decade – followed by a see-saw pattern of mild increases in the 1990s and declines in the 2000’s.

Figure 4: Average Wage Decomposition

Notes: Data from the CPS and the NLRB. Each series is represented as a log change from the corresponding 1980 level. Wage data is from both union and nonunion workers and is adjusted for worker characteristics. The 'Observed' wage series represents the national average of city-industry wages using the size of the city-industry in 1980 as fixed weights. 'Fixed Union Proportion' holds the proportion of union workers fixed at the 1980 levels. 'Full Counterfactual', in addition, holds the threat, union premium and union transitions at 1980 levels. Details of series construction described in main text.

In our decomposition, we hold the driving forces related to the de-unionisation impact

37 In particular, mean wages correspond to the wages of white workers, holding the proportion of education × gender groups at their 1980 levels.

38 We end our figure in 2010 because we only have data on one element of our decomposition – the part related to union elections - up to that year.
channels constant at their 1980 levels while allowing all other factors determining wages to vary. Our first component corresponds to the first channel: holding the the union proportions, $P_{ct}$, in (24) constant at their 1980 values. It is given by the ‘Fixed Union Proportion’ line in the figure. According to that line, this channel for the impact of the unionisation decline generated a 2.0% drop in the mean wage in the 1980s, accounting for about 13% of the overall drop in the mean wage in that decade, with about the same effect on the drop between 1980 and 2010.

In the second step in our decomposition, we construct counterfactual outside option values for both nonunion and union workers in each $ic$ cell by setting $P(\psi_{icft}, \lambda_{ct}), T_{nct}^{n}, T_{uvt}^{u}, (\eta_{ict}^{n} - \eta_{ict}^{u})$ and $(w_{nt}^{n} - w_{nt}^{u})$ equal to their 1980 values. Notice that we hold constant differences in industrial structure and wages between union and nonunion sectors, allowing the outside option values to move with movements in the $\eta_{ict}$’s and $w_{nt}$. We use the counterfactual outside option values to predict mean nonunion and union wages in each $ic$ cell and year using the estimated coefficients from our preferred specification in column (3) of Table 3. These initial estimated wages, however, are only first round effects of de-unionisation. If these counterfactual wages in a particular $ic$ cell are higher than what was actually observed then the outside options for other workers would also be higher and, so, we create a second round of counterfactual values of the outside options in which we use the first round counterfactual wages and then form a second round of counterfactual wages using the updated outside options. This new set of wages, in turn, implies a new set of outside options and, therefore, new wages. We continue to iterate on this process until the predicted wages change by less than 0.1 of a percent. This provides estimates of the effect of reductions in the threat of unionisation that reflects the complete feedback loop inherent in bargaining schemes. It also means that the union premia used in the outside option terms are consistent with the premia we would calculate from the set of counterfactual wages.

The top line in the figure shows the combination of the two steps in the decomposition – the full counterfactual effect of holding union related factors at their 1980 values. For the first decade – 1980 to 1990 – there is no spillover effect of de-unionisation. Instead, the effects are completely captured by the standard ‘between’ component in a shift-share analysis. But in the ensuing decades, spillovers begin to emerge and for the full period from 1980 to 2010, amount to almost twice the simple ‘between’ component (accounting for a 0.037 decline in the overall wage compared to 0.02 from the union proportion component). Over this full period, the two components together imply that de-unionisation can account for 37.6% of the total decline in the mean wage over that period. This is similar to Fortin et al. (2021)’s

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40 In particular, we construct an initial counterfactual wage for each nonunion worker in 1990 by subtracting the bargaining effect variable $\left(1 - P(\psi_{icf,1990}, \lambda_{c}) \cdot E_{ic,1990}\right)$ times its estimated coefficient (0.70) from their 1990 wage then adding back $0.7 \cdot \left(1 - P(\psi_{icf,1980}, \lambda_{c}) \cdot E_{ic,1980}\right)$, where $P(\psi_{icf,1980})$ is the relevant probability of facing a union election in 1980 and $E_{ic,1980}$ is the outside option value in 1980. We do the same for union wages, using our estimate of $\gamma_{1}^{u}$ from Table 3, column 3 (0.63).

41 Amazon touted its recent move to a $15 minimum wage for its workers as a reason for workers not to unionise at one of its plants. Source: Amazon to hike wages for over 500,000 workers. Our procedure would capture that initial wage increase as a standard threat effect but would also take account of the possibility that the increased wage at Amazon represents an improved outside option for other nonunion workers, leading to increases in their wages.
finding that taking account of spillovers roughly doubles the estimated ‘shift-share’ impact of de-unionisation on wage inequality over the 1979-2017 period (though, in our case, it much more than doubles it). Our results focus on wage levels rather than inequality and provide an identification strategy for supporting a claim that we have estimated causal effects.

We will examine why there was no spillover effect from 1980 to 1990 in the next section when we examine nonunion wage movements on their own. First, though, we show our counterfactual exercise for various sub-groups in the top panel of Table 5.2. Men experienced a decline in the mean real wage between 1980 and 2010 that was over double that experienced by women, but with identical sized spillover effects, deunionisation plays a larger role in explaining the decline for women than men (43% for women versus 27% for men). Older aged workers (age 36 to 55) experienced larger real wage declines than for younger workers (age 20 to 26) but the union effects are similar in levels for both groups, implying they are proportionally more important for younger workers. The combination of high school graduates and drop-outs faced real wage declines that were almost triple those with post-secondary education. Perhaps not surprisingly, the effect of de-unionisation was also much larger for the lower educated group, accounting for 43% of their real wage decline. For the post-secondary educated, the spillover effects actually imply increases in mean wages. This arises because union jobs for this education group became more concentrated in higher paying (public sector) jobs, implying increased average union wages that more than offset declines in the probability of getting a union job in the calculation of their outside option term.

<table>
<thead>
<tr>
<th></th>
<th>(1) Observed</th>
<th>(2) Union Prop.</th>
<th>(3) Full Counterfactual</th>
<th>(4) Sex: Age: Education</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>All</td>
<td>Men</td>
<td>Women</td>
<td>Age 36-55: Age 20-26: ≤ HS: &gt; HS</td>
</tr>
<tr>
<td>1980-2010</td>
<td>–0.152</td>
<td>–0.204</td>
<td>–0.088</td>
<td>–0.271: –0.189: –0.226: –0.084</td>
</tr>
<tr>
<td>(1) Observed</td>
<td>–0.152</td>
<td>–0.204</td>
<td>–0.088</td>
<td>–0.271: –0.189: –0.226: –0.084</td>
</tr>
<tr>
<td>(2) Union Prop.</td>
<td>0.020</td>
<td>0.027</td>
<td>0.011</td>
<td>0.019: 0.026: 0.043: 0.010</td>
</tr>
<tr>
<td>(3) Full Counterfactual</td>
<td>0.037</td>
<td>0.027</td>
<td>0.027</td>
<td>0.031: 0.026: 0.054: –0.019</td>
</tr>
</tbody>
</table>

Non-union:

| (4) Full Counterfactual | 0.018 | 0.015 | 0.015 | 0.014: 0.016: 0.029: –0.004 |
| (4a) Fixed Transitions  | 0.021 | 0.019 | 0.012 | 0.021: 0.020: 0.037: 0.003  |
| (4b) Fixed Union Prem.  | –0.002 | –0.003 | 0.004 | –0.005: –0.003: –0.007: –0.006 |
| (4c) Threat Effect      | –0.001 | –0.001 | –0.001 | –0.002: –0.001: –0.001: –0.002 |

Union:

| (5) Full Counterfactual | 0.019 | 0.013 | 0.012 | 0.017: 0.010: 0.025: –0.015 |

| (6) Total              | 0.057 | 0.055 | 0.038 | 0.050: 0.052: 0.097: –0.009 |
| (7) Total/Observed     | –0.376 | –0.268 | –0.429 | –0.183: –0.276: –0.432: 0.112 |

**Notes:** This table displays results from the decomposition for union and nonunion workers from 1980-2010. Each column contains the decomposition results for a different subsample. All figures are log changes from 1980 levels. Details described in main text.
6 Decomposing Nonunion Wages

We next turn to decomposing movements in mean nonunion wages, which, of course, are the focus of our estimation. We do this with a similar approach, holding constant at their 1980 values (in turn) $P(\psi_{icf}, \lambda_c)$ (the threat effect), $T^n_{jc}$ plus $(\eta^{n}_{ict} - \eta^{u}_{ict})$ (the effect of reduced probabilities of finding union jobs), and $(w^u - w^n)$ (the effect of changes in the union wage premium). As with the full decomposition, we incorporate subsequent changes in outside option values through an iterative process. Unlike the full decomposition, the result is not completely consistent. We do not update union wages in the iterative process, keeping them at the values in the specific year. Nonetheless, we view the result as useful in showing the broad strokes of which elements of de-unionisation had the most salient spillover effects onto nonunion wages.

In figure 5 we present the movements in the different union-related components of the nonunion mean wage decomposition. The line with square symbols shows the impact of holding the probability of a firm facing a unionisation drive constant at 1980 values. That impact is small, which arises because the threat probability was quite small even in 1980. Thus, while our estimates show clear evidence of the standard emulation threat effect, their actual impact on nonunion wage movements was small. That means that the sizeable spillover effect that emerges by 2010 in figure 4 is almost completely accounted for by the bargaining channel. This has potentially important implications for policy making aimed at raising wages since the threat effect can only be harness by increasing unionisation. But the bargaining channel is not unique to unions - any policy that pushes up the outside option value for workers (such as eliminating non-compete clauses or expanding commuting options) can have this effect and our results imply that this channel can be powerful. This is reminiscent of the results in, who show that wages are increasing in their index of the value of outside options. Their index increases when workers have greater probabilities of transferring to other occupations and job opportunities. Our result is driven by decreases in the probability a worker can transfer to a union job.

But why was this channel so silent during the 1980s – the decade of US unionisation’s biggest collapse? The answer can be seen in two offsetting impacts on outside option values revealed in Figure 5. On one side is the predicted impact from holding the probability a nonunion worker can move to a union job constant (the combination of $T^n_{jc}$ and $(\eta^{n}_{ict} - \eta^{u}_{ict})$), which is reflected in the ‘Fixed Transitions’ line in the figure. This line indicates that nonunion wages would have been 0.023 log points higher in 1990 and 0.016 log points higher in 2010 if these transition probabilities had not declined, accounting for between 10 and 15 percent of the declines in nonunion wages over this period. These effects are similar in magnitude to the standard ‘between’ effect due to the changes in $P_c$ in the first step of the overall mean wage decomposition. But over the 1980s, they were nearly offset by the impact of the increase in the union wage premium that we saw in figure 3. Thus, the lack of a spillover effect in that decade in the overall decomposition appears to arise from the changes in the union environment having little impact on outside option values as declining probabilities of getting a union job were balanced with higher wage premia if a worker did manage to get one. As we described earlier, our model provides an explanation for why

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42Note that the components in the nonunion wage decomposition will not add up to what is observed in
union premia would increase exactly when union power is being most substantially reduced. It stems from the reduction in the need for some nonunion firms to emulate union wages since they no longer fear their shop being unionised.

![Graph showing decomposition components: Nonunion workers](image)

**Figure 5: Decomposition components: Nonunion workers**

- **Notes:** Data from the CPS and the NLRB. Each series is represented as a log change from the corresponding 1980 level. Wage data is for nonunion workers and is adjusted for worker characteristics. Each series corresponds to a component of the decomposition, described in main text.

The second panel in Table 5.2 shows the decomposition for nonunion wages for the full 1980-2010 sample period for different sub-groups. As with the decomposition of the overall wage, the union counterfactual effect is similar in size between men and women and across age groups. Also similarly to the overall wage decomposition, the implied deunionisation effect is much larger for high school or less educated workers than for workers with a post-secondary education. The nonunion wage decomposition indicates that this effect is due to the reduced probability of the lower educated nonunion workers moving to union jobs.

### 7 Conclusion

In this paper we provide new estimates of the impact of unions on nonunion wage setting. We allow the presence of unions to affect nonunion wages both through the typically discussed channel of nonunion firms emulating union wages in order to fend off the threat of the full decomposition in figure 4 because they do not include impacts on union wages and union wages are not updated in the iteration process in the nonunion decomposition.
unionisation and through a bargaining channel in which nonunion workers use the presence of union jobs as part of their outside option. We specify these channels in a search and bargaining model that includes union formation and, in our most complete model, the possibility of nonunion firm responses to the threat of unionisation. By formalising wage setting and union formation we derive a specification grounded in theory that provides guidance on what to control for, how to interpret our coefficients and what is in the error term. Based on that, we derive a set of instruments and a model-based over-identification test, the values for which imply that our identification strategy is appropriate for this data.

Our estimates indicate that deunionisation in the US after 1980 had a substantial effect on nonunion wages, in particular, and the wage structure in general. In a decomposition exercise, holding the probability a worker can find a union job, the probability a firm faces a unionisation drive, and union wage premia constant at their 1980 levels would have undone 38% of the 16% decline in the mean (composition constant) real wage in a typical city in the US between 1980 and 2010. While we find evidence for spillover effects of unions onto nonunion wage setting through both the traditional threat channel and the bargaining channel, it is the latter that dominates. That is important for policy makers looking for tools to help in raising wages. The union threat channel can only be implemented by increasing union power. But the bargaining channel is not specific to unions. Any policy that raises worker outside option values will raise wages for a wide set of workers (?, ?). Unions are just one mechanism for doing that – though our estimates indicate a powerful and direct one.