How Do Firing Costs Affect Worker Flows in a World with Adverse Selection?

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This article provides theoretical and empirical analyses of a firing costs model with adverse selection. Our theory suggests that, as firing costs increase, firms increasingly prefer hiring employed workers, who are less likely to be lemons. Estimates of re-employment probabilities from the National Longitudinal Survey of Youth support this prediction. Unjust-dismissal provisions in U.S. states reduce the re-employment probabilities of unemployed workers relative to employed workers. Consistent with a lemons story, the relative effects of unjust-dismissal provisions on the unemployed are generally smaller for union workers and those who lost their previous jobs due to the end of a contract.

I. Introduction

Job security regulations are widely believed to reduce firing and hiring. Reduced dismissal rates benefit employed workers, but reduced hiring

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rates hurt both the unemployed and those employed workers who would like to change jobs. The purpose of this article is to present and test a theory that suggests that the reduction in hiring caused by firing costs can affect different groups of workers very differently. We show that the reduction in hiring is likely to be more severe for the unemployed than the employed. In a world with adverse selection, firing costs not only lengthen jobless spells; they may also redistribute new employment opportunities from unemployed to employed workers.

The principal theoretical innovation in our article is the introduction of adverse selection in a model of firing costs. A standard result in theoretical discussions of firing costs is that hiring and firing rates both fall when workers are protected (see, e.g., Bentolila and Bertola 1990; and Hopenhayn and Rogerson 1993). Re-employment probabilities in models of adverse selection vary according to the type of separation (Gibbons and Katz 1991). Our setup combines these features: worker quality is imperfectly observed, and so firms must contend with the possibility of hiring a “lemon” in addition to the possible expense of dismissal. When faced with an adverse shock, firms prefer dismissing bad workers who generate lower profits. As a result, in equilibrium, employed workers are more productive than the unemployed. Thus, firms concerned about firing costs will generally find it worthwhile to recruit new workers primarily from the pool of those already employed.

The central theoretical prediction of our model is that an increase in firing costs typically increases discrimination against unemployed job seekers. To test this model, we look at the effect of state unjust-dismissal provisions on the re-employment and job transition probabilities of unemployed and employed workers, using the National Longitudinal Survey of Youth (NLSY). These data are useful for this purpose for three reasons. First, the NLSY allows us to identify job-to-job transitions. Second, we can identify employed and unemployed job seekers. Finally, the NLSY’s

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geocode file allows us to identify workers covered by unjust-dismissal provisions.

Our results show reduced re-employment probabilities for the unemployed but not for employed workers over the 1980s in states that introduced unjust-dismissal provisions. The results are unchanged when controlling for state and time effects, for time-specific and state-specific effects on the unemployed, for interactions between region and time effects, for unemployment benefit receipt, for differential welfare benefits across states, and for differential effects of other variables on unemployed and employed workers. Finally, we check if the relative effect on the unemployed is greater for nonunionized than for unionized workers, since unionized jobs are covered by layoff-by-seniority rules, and for dismissed workers than for workers who are unemployed because of the end of a contract. As the lemons story suggests, our empirical findings generally confirm greater effects on nonunionized than on unionized workers and on dismissed workers than on end-of-contract workers.

As indicated above, our article relates to the literatures on adverse selection and firing costs. We extend the influential work on adverse selection by Akerlof (1970) and Spence (1973) and that on adverse selection in the labor market by Gibbons and Katz (1991) by including firing costs. More recent asymmetric information models also appear in papers by Montgomery (1999), Strand (2000), and Canziani and Petrongolo (2001). In contrast to our article, these papers consider re-employment transitions only for the unemployed and not for employed workers.

This article also relates to the extensive literature that examines the link between firing costs and labor market performance. In contrast to firing costs models without adverse selection, our model predicts a shift in hiring from unemployed to employed workers. Moreover, our work relates to the extensive empirical literature on the impact of firing costs using macrodata and microdata. Our article shares the methodology with papers using microdata and the state variation in unjust-dismissal provisions exploited by Dertouzos and Karoly (1992), Hamermesh (1993), Miles (2000), and Autor (2003). Finally, we contribute to the literature contrasting job search outcomes for employed and unemployed workers (e.g., Holzer 1987; Blau and Robins 1990).

The article is organized as follows. Section II presents and solves the matching model with asymmetric information and on-the-job search. Section III extends this model to allow for endogenous meeting rates. In

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1 Studies using aggregate data include Bertola (1990), Lazear (1990), and DiTella and MacCulloch (forthcoming), among others. There are also a handful of studies examining the impact of firing costs using microdata, including Kugler (1999, forthcoming), Oyer and Schaefer (2000, 2002), Acemoglu and Angrist (2001), Kugler, Jimeno, and Hernanz (2003), and Kugler and Pica (2004).
Section IV, we describe the data and present estimates of the effect of dismissal costs on re-employment and job transition probabilities of unemployed and employed workers. We conclude in Section V.

II. The Model
A. Description of the Model
Our theoretical framework is based on Mortensen and Pissarides (1994), where we simplify some aspects to preserve tractability but introduce dismissal costs, on-the-job search, and adverse selection to capture the phenomena of interest. We show that, in contrast to a world without adverse selection, firing costs not only reduce hiring and firing but could also shift hiring from unemployed to employed workers.

The total labor force is split between two types of workers, “good” and “bad.” The proportion of good workers is denoted by $z$, where the total labor force is normalized to one. Prior to hiring, firms observe neither the quality of applicants nor their past labor history. They only observe the applicant’s current employment status. Immediately after hiring, however, firms observe the productivity of the worker.\footnote{This assumption is made for simplicity, as it reduces the number of individual states one has to keep track of. One could specify a learning process about the worker’s productivity (as in Jovanovic 1979), but, since we are not dealing with learning aspects, we keep this part of the model as simple as possible.}

Firms freely enter the market by creating vacant positions. Once the position is created, firms face a cost $C$ of holding a vacancy. Because of free entry, the value of a vacancy must always be equal to zero in equilibrium. A job seeker meets a vacant job with probability $a$ per unit of time, which we take to be exogenous in this section but endogenize in the following section. When meeting a worker, a firm decides whether to hire a worker or not, depending on his labor market status. Below we only consider the case where employed job seekers are hired for sure, which must hold in any equilibrium of interest.\footnote{As will be made clear below, the pool of employed job seekers is of higher quality than that of unemployed job seekers, thus generating higher net expected profits. If firms did not hire employed applicants with probability one, then these profits would be negative or zero, implying strictly negative net expected profits out of unemployed applicants. The exit rate from unemployment would then be zero, and so the whole workforce would be unemployed in equilibrium.}

Once a position is filled, production takes place. The firm’s output per unit of time is $m + \eta$, where $m$ is a match-specific component and $\eta$ is worker specific. We assume that $\eta = \eta_g$ for good workers and $\eta = \eta_b$ for bad workers. When the match is initially formed, the match-specific component is equal to but, with probability $\gamma$ per unit of time, the firm is hit by a shock that changes the productivity of the match.
Every time such a shock occurs, the new productivity is drawn from a distribution \( G(m) \) over the interval \([\eta, \bar{m}]\).

Wages are assumed to be equal to a base wage, \( \bar{\omega} \), plus a constant fraction, \( \varphi \), of output with worker productivity, \( \eta \), and match productivity, \( m \):

\[
\omega(m, \eta) = \varphi(m + \eta) + (1 - \varphi)\bar{\omega}.
\]

This assumption implies that firms make higher profits out of good workers than out of bad ones, which is central to our results.\(^4\)

Production takes place until either the firm closes the position or the worker quits. Firms pay a tax \( F \) when dismissing a worker, which is paid to a third party. We specify firing costs as a tax because, as discussed below, a substantial fraction of firing costs go to third parties such as lawyers, insurers, and the government. In contrast, the firm does not have to pay \( F \) when the worker quits. The quit rate is endogenous and given by the probability of engaging in on-the-job search times the instantaneous probability of receiving an offer, \( a \). Workers face a flow search cost, \( c \), from searching on the job, but the benefit of searching is that they move to a match with the highest possible level of productivity. Whether the firm dismisses the worker or the worker quits, the position is destroyed and the firm’s value drops to zero.

B. Equilibrium

1. On-the-Job Search

We first solve for on-the-job search, which is obtained by comparing the worker’s value of being employed while searching and not searching. Let \( E(m, \eta, NS) \) be the value of being employed while not searching for an employed worker of productivity \( \eta \) and match-specific productivity \( m \). The value of the employed worker who does not search is given by the following Bellman equation:

\[
rE(m, \eta, NS) = \varphi(m + \eta) + (1 - \varphi)\bar{\omega}
\]

\[
+ \gamma \int_{m_c}^{\bar{m}} E(m, \eta, g(m)) dm + G(m, \eta) U(\eta)
\]

\[
- E(m, \eta, NS),
\]

where \( U(\eta) \) is the value of being unemployed and \( m_c \) is the critical value of the match-specific productivity that triggers a dismissal, and so the last

\(^4\)This holds in a wide variety of models of wage formation (except for the perfectly competitive case, which equates wages to marginal product).
term is the expected capital gain or loss from being hit by a shock. Similarly, the value of an employed job seeker is given by the following Bellman equation:

\[
\bar{r}E(m, \eta, S) = \varphi(m + \eta) + (1 - \varphi)\bar{w} - c + a|E(\bar{m}, \eta) - E(m, \eta, S)|
\]

\[ + \gamma \int_{m_c}^{\bar{m}} E(m, \eta)g(m)dm + G(m(\eta))U(\eta) \]

\[ - E(m, \eta, S), \]

where the fourth term represents the expected capital gain from quitting into a new job.

Search while on the job for an employed worker with match-specific productivity, \( m \), takes place if \( E(m, \eta, S) \geq E(m, \eta, NS) \). Since the cost of search is constant and the benefit from searching is that the person moves from the current match to the highest possible match-specific productivity, the gains from searching while on the job increase as the current match level decreases. This means that on-the-job search is given up at the unique value, \( \bar{m} \), below which there is always on-the-job search, and which satisfies the condition

\[ E(\bar{m}, \eta, S) = E(\bar{m}, \eta, NS). \]

Substituting (1) and (2) into the above condition, we can solve for \( \bar{m} \) to obtain

\[ \bar{m} = \bar{m} - \frac{(r + \gamma)c}{\varphi a}, \]

which means that search behavior is independent of worker type. Since the case of interest is given by the case where some workers engage in search, we limit ourselves to the case where the search threshold exceeds the dismissal threshold, that is, \( \bar{m} > m(\eta) \), for one or both type of workers.\(^5\)

2. Firing and Hiring Decisions

Given that the residual value of firing the worker is zero, the firm fires the worker if \( J(m, \eta) < -F \). The value of a job filled with an employed

\(^5\) Sufficient conditions for the search thresholds to exceed the dismissal thresholds are given in n. 7.
job seeker of productivity $\eta$ and match-specific productivity $m \leq \bar{m}$, $f(m, \eta)$, is given by the following Bellman equation:

$$r f(m, \eta) = (1 - \phi)(m + \eta - \bar{w}) - a f(m, \eta)$$

$$+ \gamma \left[ \int_{m_i}^m f(m, \eta) g(m) dm - G(m, \eta) F - f(m, \eta) \right].$$

The second term on the right-hand side is the expected capital loss experienced by the firm if the worker quits, which, conditional on on-the-job search, happens with instantaneous probability $a$, and the last term is the expected capital gain or loss associated with a productivity shock. Solving for $f(m, \eta)$, we obtain

$$f(m, \eta) = \frac{\gamma \hat{f}(\eta) + (1 - \phi)(m + \eta - \bar{w})}{(r + \gamma + a)},$$

where $\hat{f}(\eta) = \int_{m_i}^m f(m, \eta) g(m) dm - G(m, \eta) F$ is the average value of the match to the firm over the current value of the shock. Similarly, the value of a job filled with an employed worker who does not search, that is, $m > \bar{m}$, is

$$f(m, \eta) = \frac{\gamma \hat{f}(\eta) + (1 - \phi)(m + \eta - \bar{w})}{(r + \gamma)}.$$

Since we consider the case in which some workers search before reaching the dismissal threshold, that is, $m_i(\eta) \leq \bar{m}$, and the dismissal threshold is given by $f(m_i(\eta), \eta) = -F$, we obtain the following solution for the dismissal threshold:

$$m_i(\eta) = -F(r + \gamma + a) - (1 - \phi)\eta + (1 - \phi)\bar{w} - \gamma \hat{f}(\eta)$$

$$= \frac{-(r + \gamma + a) \hat{f}(\eta) + \gamma \hat{f}(\eta) + (1 - \phi)(m + \eta - \bar{w})}{(1 - \phi)},$$

which defines a relationship between $m_i(\eta)$ and $\hat{f}(\eta)$. Substituting (3) and (4) into $\hat{f}(\eta)$ provides the other relationship between these two unknowns,

$$\hat{f}(\eta) = \frac{\int_a^\infty \frac{(1 - \phi)(m + \eta - \bar{w})}{(r + \gamma)} g(m) dm + \int_a^\infty \frac{(1 - \phi)(m + \eta - \bar{w})}{(r + \gamma + a)} g(m) dm - FG(m, \eta)}{1 - \frac{\gamma}{(r + \gamma)} [G(\bar{m}) - G(\bar{m})] - \frac{\gamma}{(r + \gamma + a)} [G(\bar{m}) - G(m, \eta)]}.$$  

Without on-the-job search, this relationship is given by

$$m_i(\eta) = \frac{-F(r + \gamma) - (1 - \phi)\eta + (1 - \phi)\bar{w} - \gamma \hat{f}(\eta)}{(1 - \phi)},$$

indicating that on-the-job search lowers the dismissal threshold. As in Saint-Paul (1995), firms faced with firing costs may prefer to use attrition and wait until workers quit rather than dismiss at a cost, $F$. 


and substituting this equation into (4) determines \( m_1(\eta) \) uniquely.

Furthermore, \( m_1 \) is falling with \( \eta \), falling with \( F \), and increasing with \( \tilde{w} \). This means that \( m_1(\eta_L) > m_1(\eta_H) \), and so we may have two cases with on-the-job search: (a) only some workers search, that is, \( m_1(\eta_L) > \tilde{m} \geq m_1(\eta_H) \), in which case only good workers search, or (b) both types of workers search, that is, \( \tilde{m} \geq m_1(\eta_L) > m_1(\eta_H) \). We consider the second case, which is the least restrictive one, although our results below are strengthened under the first case.\(^7\) Moreover, the dismissal threshold of good workers is more responsive to changes in \( F \) and than the dismissal threshold of bad workers. Consequently, 

\[
\frac{dm_1(\eta_L)}{dF} < \frac{dm_1(\eta_H)}{dF} \quad \text{and} \quad \frac{dm_1(\eta_L)}{d\tilde{w}} > \frac{dm_1(\eta_H)}{d\tilde{w}},
\]

which is due to a discount effect, since good workers are less likely to be fired (see the appendix for the proof).

We now compute the hiring decision of a firm faced with an applicant. The quality of the applicant is unobservable, but his status is observable and provides a signal to the firm. Let \( z_g \) and \( z_b \) be the proportion of good workers among, respectively, the employed and unemployed job seekers. Then the expected present discounted values associated with hiring an employed and an unemployed applicant are, respectively,

\[
\Pi_E = z_g J(\tilde{m}, \eta_L) + (1 - z_g) J(\tilde{m}, \eta_H), \tag{5}
\]

\[
\Pi_u = z_b J(\tilde{m}, \eta_H) + (1 - z_b) J(\tilde{m}, \eta_L). \tag{6}
\]

Firms prefer to hire an employed applicant rather than an unemployed one, that is, \( \Pi_E > \Pi_u \), since good workers are dismissed less often than bad ones, and, thus, \( z_g > z_b \), and firms make more profits out of good workers, \( J(\tilde{m}, \eta_L) > J(\tilde{m}, \eta_H) \).

The firm hires the worker if \( \Pi_E > 0 \), it does not hire if \( \Pi_E < 0 \), and it is indifferent if \( \Pi_E = 0 \). Letting \( p_u \) be the probability that an unemployed worker is hired, then the hiring behavior is represented in the \((p_u, z_u)\) plane by the economic behavior \((EB)\) locus in figure 1. If the economy

\[
\frac{(r + \gamma)c}{\varphi a} \leq \tilde{m} + \frac{F(r + 2\gamma)}{(1 - \varphi)} + \eta_L - \tilde{\omega}.
\]

A sufficient condition for this to hold is

\[
\frac{(r + \gamma)c}{\varphi a} \leq \tilde{m} + \frac{F(r + 2\gamma)}{(1 - \varphi)} + \eta_L - \tilde{w}.
\]

Similarly, a sufficient condition for good workers to engage in on-the-job search is

\[
\frac{(r + \gamma)c}{\varphi a} \leq \tilde{m} + \frac{F(r + 2\gamma)}{(1 - \varphi)} + \eta_L - \tilde{w}.
\]
Firing Costs and Worker Flows

Fig. 1.—Economic behavior locus

is above the horizontal $PP$ line, then $\Pi_s > 0$. In this case, all unemployed and employed applicants are hired. If the economy lies on $PP$, then $\Pi_s > \Pi_u = 0$. In this case, all employed applicants are hired, while unemployed applicants are only hired with probability $p_u$, and, thus, there is discrimination in hiring against the unemployed. The lower hiring rate of the unemployed relative to employed workers reflects statistical discrimination, since firms use employment status to predict productivity. If the economy lies on $PP$, the $EB$ locus shifts upward whenever one of the labor cost parameters, $F$ or $\bar{w}$, increases. This is because any parameter change that reduces profits must be offset by an increase in the quality of unemployed applicants. Otherwise, the incentive to hire them would disappear (see the appendix for the proof).

3. Steady State Analysis

We complete the joint determination of $p_u$ and $z_u$ by deriving a steady state relationship between the two. In steady state, inflows into unemployment must be equal to outflows for each group of workers. Letting $u$ be the unemployment rate, then

$$\gamma G_{H}(z - u z_u) = a_p u z_u,$$

$$\gamma G_{L}(1 - z - u(1 - z_u)) = a_p u (1 - z_u),$$

where $G_i = G(m_i; \eta_i)$, $i = H, L$, and $l_i = (z - u z_u)$ is total employment of good workers and $l_i = [1 - z - u (1 - z_u)]$ is total employment of bad

8 We ignore the case where $\Pi_s < 0$, as it would imply a zero outflow from unemployment, and therefore all workers would be unemployed in steady state.
workers. Eliminating $u$ between these two equations allows us to derive the following steady state relationship between $p_u$ and $z_u$,

$$z_u = z \frac{\gamma + ap_u}{G_l}. \quad (9)$$

This equation determines the steady state (S-S) locus, which provides a condition between $p_u$ and $z_u$ that keeps the composition of employment and unemployment time invariant. The S-S locus is downward sloping because a lower $p_u$, that is, a lower exit rate from unemployment, makes the steady state composition of the unemployment pool more similar to its source population, that is, the employed, who are of better quality.

The equilibrium is determined by the point where the S-S locus crosses the $EB$ locus. Hiring discrimination against the unemployed (i.e., $p_u < 1$) arises whenever the S-S locus cuts the $EB$ locus along its horizontal portion $PP$ (see fig. 2a). The job finding rates of employed and unemployed job seekers are the same if the S-S locus cuts the $EB$ locus from above (see fig. 2b).

The total number of employed job seekers is given by the fraction of good workers who search among all good workers plus the fraction of bad workers who search among all bad workers,

$$l_e = l_{ht} \left( \frac{\tilde{G} - G_{ht}}{1 - G_{ht}} \right) + l_i \left( \frac{\tilde{G} - G_{li}}{1 - G_{li}} \right).$$

Consequently, using the steady state conditions, the proportion of good workers among employed job seekers is

$$z_e = \frac{l_{ht} \left( \frac{\tilde{G} - G_{ht}}{1 - G_{ht}} \right)}{l_{ht} \left( \frac{\tilde{G} - G_{ht}}{1 - G_{ht}} \right) + l_i \left( \frac{\tilde{G} - G_{li}}{1 - G_{li}} \right)}.$$

where $\tilde{G} = G(\tilde{m})$.

C. Effects of Firing Costs

The comparative statics relationship of interest is the effect of changes in firing costs, $F$, on the hiring of the unemployed. We have already seen that the $EB$ locus shifts upward when $F$ increases. If the S-S locus did not move, the increase in firing costs would make firms choosier and
Fig. 2.—Equilibria:  

a, Equilibrium with $1 > \rho_u^* > 0$.  
b, Equilibrium with $\rho_u^* = 1$.  
c, Equilibrium with $\rho_u^* = 0$.  

$z_u^* = \frac{1}{2} z_u$
would lead to greater discrimination of the unemployed, that is, to a fall in \( p_u \). However, the S-S locus does move when \( F \) increases, because it affects the firing margins \( m_u(\eta_u) \) and \( m_l(\eta_l) \). Both of the inflows into unemployment of good and bad workers are reduced by firing costs. If the latter were reduced more than the former, then the quality of the unemployed would improve. However, if the inflow of bad workers were reduced more than the inflow of good workers, then the quality of the unemployed would worsen. The S-S locus could then move up or down. Proposition 1 shows that, under reasonable conditions about the distribution of the shocks, the S-S locus shifts down so that an increase in \( F \) unambiguously reduces \( p_u \).

**Proposition 1.** If the distribution \( G \) satisfies the nonincreasing hazard property, that is,

\[
\frac{g(m)}{G(m)} \text{ is nonincreasing with } m,
\]

the S-S locus moves down when firing costs, \( F \), increase.

Proposition 1 (see the appendix for the proof) shows that, given hiring policies, an increase in firing costs decreases the job loss rate more for good workers than for bad workers and, therefore, worsens the quality of the unemployed. This comes from two effects. First, as shown above, the dismissal threshold is more sensitive for good workers because of lower discounting. Second, if the non-increasing hazards assumption holds, a given change in the dismissal threshold has a greater relative effect on the number of people being fired the smaller that number. Since fewer good workers are fired, their firing rate falls proportionately more than for bad workers, thus reducing the average quality of job losers.

Figure 3 illustrates proposition 1. Higher firing costs exacerbate discrimination in hiring against unemployed workers, both because firms require better unemployed applicants and because under (10) firing costs reduce the quality of job losers. In this model with adverse selection, higher firing costs are likely to reduce job transition probabilities for unemployed but not for employed workers, for whom it is given by the exogenous instantaneous probability of receiving an offer, \( a \). When \( a \) is made endogenous, as will be done in the next section, the job finding

\[
9 \text{ Of course, the non-increasing hazards assumption need not hold, but it holds for a wide range of distributions, including the uniform distribution and any distribution that does not have an accentuated interior mode. Even when property (10) does not hold, we can still establish that, under reasonable conditions, discrimination against the unemployed disappears with low enough firing costs (see the appendix).}
\]

\[
10 \text{ In contrast, while an increase in wages also shifts the } EB \text{ locus up, its effect on the S-S locus goes in the opposite direction of the effect of firing costs when the non-increasing hazard assumption holds.}
\]
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rate of employed job seekers may also fall with $F$, but by less than the unemployeds’ rate. Furthermore, as shown below, the job-to-job flow may even go up with $F$ as a greater fraction of the employed engage in on-the-job search.

III. Endogenous Meeting Rates

In this section, the meeting rate between firms and workers is now determined by firms’ optimal choices about vacancies. In particular, total
contacts between searching firms and workers are generated by a matching function,

\[ b = m(v, u + l), \]

where \( v \) is the number of vacant jobs. Consequently, the arrival rate of offers and the arrival rates of employed and unemployed job seekers are

\[ a = \frac{m(v, u + l)}{u + l} = m(\theta, 1), \]

\[ \lambda_s = \frac{\frac{1}{u + l}}{v} m(v, u + l) = \frac{l}{u + l} m(1, \frac{1}{\theta}) = \frac{l}{u + l} q(a), \quad (11) \]

\[ \lambda_u = \frac{\frac{1}{u + l}}{v} m(v, u + l) = \frac{u}{u + l} m(1, \frac{1}{\theta}) = \frac{u}{u + l} q(a), \quad (12) \]

where \( \theta = v/(u + l) \) and \( q(a) = m(1, 1/\theta) \), with \( q'(a) < 0 \).

The number of vacant jobs is determined by the entry decision of firms, where the value of a vacancy \( V \) satisfies

\[ rV = -C + \lambda_s (\Pi_s - V) + \lambda_u p_s (\Pi_u - V). \quad (13) \]

An equilibrium is a set of endogenous variables such that, in addition to the equilibrium conditions derived in the previous section, the equilibrium condition holds. Hence, eliminating \( \lambda_s \) and \( \lambda_u \), we have one additional endogenous variable, \( a \), and one additional equilibrium condition.

In equilibrium, free entry implies that \( V = 0 \), and this free entry condition determines the total number of vacancies, \( v \). Given the equilibrium value of \( a \), the arrival rates and the number of vacancies can then be recursively computed using the preceding equations.

The following theorem (see the proof in the appendix) tells us that an equilibrium always exists and that, when nontrivial, it is “well behaved.”

**Theorem 1.**

i) Either there exists a zero employment equilibrium such that \( a = 0 \) and \( u = 1 \), or

ii) there exists an equilibrium with \( a > 0 \) that is “stable” in the sense that the value of the firm is locally decreasing in \( a \).

A sufficient condition for i to be ruled out is \( zf(m, \eta m) + (1 - z) \times f(m, \eta u) > 0 \), where the \( f \)'s are computed using \( a = 0 \). This means that \( a = 0 \) cannot be an equilibrium, since it would then be profitable for an atomistic firm to deviate by hiring an unemployed worker, who would never quit and is good with probability \( z \). The last part of claim ii refers to the case where \( a \) is treated as exogenous, as in the previous
Table 1
Numerical Simulations of Effects of Firing Costs on Labor Market Flows

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<td>.716</td>
<td>.262</td>
<td>.042</td>
<td>.364</td>
<td>.574</td>
<td>.056</td>
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<tr>
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<td>.698</td>
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<td>.049</td>
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<td>.691</td>
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<td>.691</td>
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<td>.679</td>
<td>.47</td>
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<td>.66</td>
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<td>.031</td>
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<td>.64</td>
<td>.13</td>
<td>.08</td>
<td>.3</td>
<td>.03</td>
<td>.347</td>
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<td>.269</td>
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<td>.627</td>
<td>.097</td>
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<td>.293</td>
<td>.029</td>
<td>.355</td>
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<td>.323</td>
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<tr>
<td>.59</td>
<td>.052</td>
<td>.03</td>
<td>.281</td>
<td>.027</td>
<td>.379</td>
<td>.658</td>
<td>.469</td>
</tr>
</tbody>
</table>

Note.—The set of parameters used for the simulation are as follows: $\gamma = .1$, $r = .55$, $m = 1$, $q = .7$, $q_1 = .5$, $\phi = 1.5$, $z = .5$, $c = 5$, $C = 2.3$, $\varphi = .5$. The $q(\cdot)$ function was chosen as $q(a) = qa^{-\rho}$, with $q_1 = 2$ and $\delta = 5$, and shocks are uniformly distributed over $[0, \bar{m}]$.

This stability property implies that any parameter shift that reduces the value of the firm given a vacancy, $V$, is negatively related to labor market tightness, $a$.

This stability property implies that any parameter shift that reduces the value of the firm given a will reduce the equilibrium value of $a$. This is typically true of an increase in the firm’s labor costs, $F$, $\bar{w}$, and $\varphi$. Since it is difficult to say more analytically, we turn to numerical simulations.

The simulation results, reported in Table 1, suggest the following conclusions:

1. For low values of $F$, low-quality workers do not engage in on-the-job search because the threshold value of $m$ at which they are fired is higher than the threshold value of $m$ below which they search. In this zone, an increase in $F$ increases $a$. In this zone, $p_u = 1$ and $z_e = 1$, as all employed job seekers are of good quality. The economy then moves to a zone where both types of workers search on the job as $F$ increases. In this zone, $p_u$ and $a$ fall monotonically with $F$, and one eventually reaches a zone where $p_u < 1$.

2. For $F > 0.1$, the unemployment-to-job flow, which is equal to $ap_u$, falls with $F$, and its fall accelerates in the zone where $p_u < 1$, where both $a$ and $p_u$ fall in response to $F$.

3. The job-to-job flow is hump-shaped as a result of two conflicting effects. First, an increase in firing costs widens the distribution of productivity levels among employed workers, thus increasing the proportion of employed workers engaged in on-the-job search. Second, it reduces $a$, thus reducing the job finding rate of all job seekers. Our simulations suggest that, at low firing costs, the first effect dominates, while the second does at high firing costs.

4. The job loss rate monotonically decreases with $F$, as expected.
5. In the zone where $p_u = 1$, the quality of unemployed job seekers falls with $F$ for $F > 0.2$. This is because, under the conditions of proposition 1, the inflow into unemployment is more responsive to $F$ for good than for bad workers. In the zone where $p_u < 1$, the quality of the unemployed goes up with $F$, due to the mechanisms already explained in the previous section. The quality of employed job seekers follows a similar U-shaped pattern.

6. The unemployment rate, $u$, falls with $F$ in the zone where $p_u = 1$, implying that the effect of a lower job loss rate is stronger than that of a lower job finding rate. The unemployment rate goes up in the zone where $p_u < 1$, where the job finding rate falls much more rapidly in response to $F$ due to the downward adjustment of the discrimination parameter $p_u$.

Similar to the results with exogenous meeting rates, these results show that, as firing costs rise, the job finding rates of the unemployed decrease relative to those of the employed.

IV. The Impact of Dismissal Costs on Accession Rates

This section provides evidence on the impact of unjust-dismissal provisions in the United States on the re-employment and job transition probabilities of unemployed and employed workers. Before turning to the data and the empirical analysis, we provide a description of the changes in firing costs in U.S. states over the 1970s and 1980s.11

A. Exceptions to Employment at Will

Until the 1959 ruling by California’s appellate court imposing restrictions on dismissals, the common-law rule known as the employment at will doctrine applied in all U.S. states. The employment-at-will doctrine determined that employers could “discharge or retain employees at will for good cause or for no cause, or even for bad cause without thereby being guilty of an unlawful act per se.”12 During the 1980s, the United States experienced a sharp and probably unanticipated rise in the recognition of exceptions to employment at will, leaving only four out of the 51 states (including the District of Columbia) as strictly employment-at-will states by the 1990s.

The exceptions to at-will employment adopted during the last few decades in the United States can be grouped into three main categories: the implied contract doctrine, the public policy doctrine, and the covenant of good faith and fair dealing. The implied contract doctrine establishes that the employment relationship may be governed by implied contractual provisions restricting the ability of employers to terminate employees.

11 Much of the description on the evolution of the legal environment that follows is based on Dertouzos and Karoly (1992), Miles (2000), and Autor (2003).
12 Payne v. Western & Atlantic Railroad, 1884, Tennessee Supreme Court, 82 Tenn. 507 (1884), 519–20, 1884 WL 469 at *6 (Sep. term 1884).
The courts establish evidence of an implied contract from written and oral statements, employment practices and manuals, employees’s length of service, and the general context of the employment relationship. While only six states had recognized the implied contract doctrine by 1981, by the 1990s, 42 states had introduced exceptions to employment at will based on the implied contract rule.

Public policy exceptions bar employers from terminating employees for refusing to commit an act contrary to public policy or for committing an act protected by public policy. The public policy doctrine is usually defined to include only statutes. Only in a few cases has it been applied more broadly to include regulations, administrative rules, and professional codes of ethics. For this reason, the public policy doctrine appears to be less restrictive to employers than the implied contract exception. The public policy doctrine was first recognized in 1959 by California, but it was only widely accepted by most states during the 1980s. By the 1990s, 42 states had recognized the public policy doctrine but only eight of them in its broader form.

A third, less widely recognized, exception is the covenant of good faith and fair dealing, which bars employers from dismissing workers in order to deprive them of earned benefits (e.g., pensions and bonuses). Many legal scholars have considered the covenant of good faith as being potentially the most far-reaching of the three doctrines, in that it can imply that dismissal must always be for cause. However, in spite of the early recognition of this doctrine in 1974, only 12 states had issued decisions in support of this exception by the 1990s.

There is little information on the actual costs imposed by these unjust-dismissal exceptions. A study of wrongful discharge cases by Dertouzos and Karouly (1992) in California, however, reported average compensatory damages of $388,500 and average legal fees of $98,000 in cases where the defense won and $220,000 in cases where the plaintiff won. The costs going to third parties (i.e., lawyers) suggest the importance of the firing tax component of dismissal costs.

In the empirical analysis that follows, we distinguish among the three types of exceptions. We construct the dismissal legislation variables using the classifications of the doctrines in Autor (2003). We checked the robustness of our results using the classification provided by Dertouzos and Karoly (1992). \[^{13}\]

\[^{13}\] Since the results using the Dertouzos and Karoly (1992) classification were almost identical to those using the classification provided by Autor (2003), we only present the results based on the latter.
B. Data Description

We use the random sample of 6,111 individuals in the NLSY for the years 1979–84 and 1996. In these years, employed workers were asked about their job search activities and, in particular, were asked whether they were looking for another job. This allows us to contrast employed and unemployed job seekers. The unemployed are defined in the usual way as individuals who did not work during the survey week and were looking for work or on layoff. Employed job seekers are defined as those individuals who worked during the survey week and were looking for work. These same definitions were used by Holzer (1987) and Blau and Robins (1990) in studies of job search by employed and unemployed workers.

The NLSY work history file allows us to track employer-specific data and to correctly identify job-to-job transitions. For multiple-job holders, the “main job” was identified as the job in which the worker earned the most during that week. Moreover, since observations are defined by search spells of employed and unemployed workers, an individual worker can contribute more than one observation if, for example, the worker is unemployed during 2 or more sample years or if the worker is an employed job seeker in one sample year and unemployed in another. We eliminated the following observations from the sample: all observations with a real wage less than one 1979 dollar, workers in the public sector, persons serving in the military, agricultural workers, and the self-employed.14 In addition, while the youngest person in the NLSY enters the sample at age 14, we restrict our sample to workers 17 years of age or older. The oldest workers reach age 39 in our sample period.

Most important for our purposes, the geocode file allows us to generate the job security provision variables, as it identifies the state of residence of each individual at the time of the interview. The geocode file contains the unemployment rate in the respondent’s county of residence.

C. Descriptive Statistics

Simple comparisons show that workers covered by unjust-dismissal provisions have higher job finding rates than do uncovered workers. This can be seen in the first row of table 2, which reports the job finding probability of workers covered by exceptions to be 0.026 higher than for uncovered workers. This difference shows the higher job finding rates of the employed covered by exceptions reported in the second column of

14 Workers in the public sector and the military are eliminated because we want to concentrate on workers hired by profit-making businesses. Agricultural workers are eliminated because these workers are likely to have seasonal contracts and unlikely to be subject to unjust-dismissal doctrines. The self-employed are eliminated because they are not subject to adverse selection problems.
Table 2

Descriptive Statistics

<table>
<thead>
<tr>
<th>Variable</th>
<th>Sample</th>
<th>Covered by Exceptions</th>
<th>Not Covered by Exceptions</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Entire Sample</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Searcher found job within a year</td>
<td>.328</td>
<td>.339</td>
<td>.313</td>
</tr>
<tr>
<td>Proportion of searchers unemployed</td>
<td>.413</td>
<td>.417</td>
<td>.406</td>
</tr>
<tr>
<td>Age</td>
<td>22.259</td>
<td>22.982</td>
<td>21.077</td>
</tr>
<tr>
<td></td>
<td>(4.556)</td>
<td>(5.075)</td>
<td>(3.324)</td>
</tr>
<tr>
<td>Male</td>
<td>.589</td>
<td>.591</td>
<td>.586</td>
</tr>
<tr>
<td>White</td>
<td>.681</td>
<td>.726</td>
<td>.616</td>
</tr>
<tr>
<td>Other race</td>
<td>.046</td>
<td>.054</td>
<td>.031</td>
</tr>
<tr>
<td>Married</td>
<td>.221</td>
<td>.239</td>
<td>.191</td>
</tr>
<tr>
<td>No. of children</td>
<td>.447</td>
<td>.535</td>
<td>.335</td>
</tr>
<tr>
<td></td>
<td>(.877)</td>
<td>(.984)</td>
<td>(.705)</td>
</tr>
<tr>
<td>Education</td>
<td>12.082</td>
<td>12.201</td>
<td>11.876</td>
</tr>
<tr>
<td></td>
<td>(1.94)</td>
<td>(1.932)</td>
<td>(1.886)</td>
</tr>
<tr>
<td>White-collar worker</td>
<td>.601</td>
<td>.609</td>
<td>.587</td>
</tr>
<tr>
<td>Manufacturing sector</td>
<td>.286</td>
<td>.281</td>
<td>.295</td>
</tr>
<tr>
<td>Union member</td>
<td>.163</td>
<td>.165</td>
<td>.159</td>
</tr>
<tr>
<td>Tenure</td>
<td>40.044</td>
<td>41.401</td>
<td>38.056</td>
</tr>
<tr>
<td></td>
<td>(25.866)</td>
<td>(26.661)</td>
<td>(24.27)</td>
</tr>
<tr>
<td>Wage</td>
<td>520.8</td>
<td>566.4</td>
<td>446.3</td>
</tr>
<tr>
<td></td>
<td>(474.9)</td>
<td>(546.3)</td>
<td>(323.7)</td>
</tr>
<tr>
<td>Nonwage income</td>
<td>16,189</td>
<td>17,757.9</td>
<td>13,705.4</td>
</tr>
<tr>
<td></td>
<td>(25,438)</td>
<td>(30,383.7)</td>
<td>(14,451.3)</td>
</tr>
<tr>
<td>Local unemployment rate</td>
<td>8.812</td>
<td>9.511</td>
<td>7.751</td>
</tr>
<tr>
<td></td>
<td>(3.599)</td>
<td>(3.829)</td>
<td>(2.932)</td>
</tr>
<tr>
<td>N</td>
<td>4,776</td>
<td>2,918</td>
<td>1,858</td>
</tr>
</tbody>
</table>

Note.—The table reports means of all variables. The sample includes only unemployed workers and employed job searchers. The “Covered” column includes means for individuals in adopting states after the adoption of the doctrines. The “Not Covered” column includes means for individuals in nonadopting states and in adopting states before adoption. Standard deviations are in parentheses.

table 3, but it hides the lower job finding rates of unemployed workers covered by exceptions reported in the first column of table 3. Other comparisons by unjust-dismissal status show broadly similar characteristics between the two groups of workers. For example, the overall proportion of searchers unemployed is 41.3% and slightly higher among those covered by exceptions than among those in employment-at-will states, but the difference is not statistically significant.\(^\text{15}\)

To illustrate the differential impact of unjust-dismissal provisions on unemployed and employed job seekers, table 3 presents a set of contrasts in accession rates between unemployed and employed individuals in adopting and nonadopting states. Panel A of table 3 shows differences in average accession rates for workers in covered states (i.e., adopting states after adoption of the doctrines) and uncovered states (i.e., nonadopting

\(^\text{15}\) This implies that about 60% of the searchers in our sample are employed. Statistics not shown in the table indicate that about 20% of employed workers search, with about equal proportions searching in covered and uncovered states.
Table 3
Average Job Finding Rates

<table>
<thead>
<tr>
<th></th>
<th>Unemployed</th>
<th>Employed</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><strong>A. All exceptions:</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Covered</td>
<td>.503</td>
<td>.22</td>
</tr>
<tr>
<td>( .01)</td>
<td>( .007)</td>
<td></td>
</tr>
<tr>
<td>Uncovered</td>
<td>.53</td>
<td>.166</td>
</tr>
<tr>
<td>( .013)</td>
<td>( .008)</td>
<td></td>
</tr>
<tr>
<td>Covered-uncovered differences</td>
<td>- .027</td>
<td>.054</td>
</tr>
<tr>
<td>( .016)</td>
<td>( .01)</td>
<td></td>
</tr>
<tr>
<td>Unemployed-employed difference in differences</td>
<td>- .081</td>
<td></td>
</tr>
<tr>
<td></td>
<td>( .019)</td>
<td></td>
</tr>
</tbody>
</table>

|                  |            |          |
| **B. Implicit contract:** |            |          |
| Covered          | .487       | .244     |
| ( .012)          | ( .009)    |          |
| Uncovered        | .531       | .17      |
| ( .01)           | ( .006)    |          |
| Covered-uncovered differences | - .044 | .073 |
| ( .016)          | ( .011)    |          |
| Unemployed-employed difference in differences | - .117 |          |
|                  | ( .019)    |          |

|                  |            |          |
| **C. Public policy:** |            |          |
| Covered          | .505       | .225     |
| ( .011)          | ( .008)    |          |
| Uncovered        | .521       | .174     |
| ( .011)          | ( .007)    |          |
| Covered-uncovered differences | - .015 | .05  |
| ( .015)          | ( .01)     |          |
| Unemployed-employed difference in differences | - .066 |          |
|                  | ( .019)    |          |

|                  |            |          |
| **D. Good faith:** |            |          |
| Covered          | .487       | .214     |
| ( .019)          | ( .013)    |          |
| Uncovered        | .519       | .194     |
| ( .009)          | ( .006)    |          |
| Covered-uncovered differences | - .032 | .02  |
| ( .021)          | ( .014)    |          |
| Unemployed-employed difference in differences | - .052 |          |
|                  | ( .025)    |          |

Note.—The first and second rows in each panel of the table report average job finding rates for unemployed and employed workers covered and not covered by unjust-dismissal doctrines. Covered workers are those living in adopting states after adoption and uncovered workers are those living in nonadopting states and in adopting states before adoption. The third and fourth rows in each panel report differences of the average job finding rates. Standard errors are in parentheses.

states as well as adopting states before adoption) for unemployed and employed workers. For example, the first and second rows in the first column show that the average job finding rate of unemployed workers in covered states is 0.503, while the average job finding rate of unemployed workers in uncovered states is 0.53. The third row presents the difference in average job finding rates between covered and uncovered workers. These results show that the average job finding rate is 0.03 lower for
unemployed workers and 0.05 higher for employed workers in adopting compared to nonadopting states. The last row in panel A contrasts the difference in average job finding rates between covered and uncovered workers for those employed and unemployed. The results show that the employed-unemployed difference in average job finding rates increased for workers living in states that introduced unjust-dismissal exceptions relative to the control group living in nonadopting states.

Panels B, C, and D of table 3 show similar estimates for unemployed and employed workers, where coverage status is defined for the implied contract, public policy, and good faith doctrines separately. The results in the last row of panels B, C, and D show that the employed-unemployed job finding difference increased in those states that introduced each of these doctrines as compared to those that did not.

D. Probit Estimates

To estimate the impact of unjust-dismissal exceptions on the re-employment probabilities of unemployed workers relative to employed workers while controlling for other variables, we estimate the following reduced-form probit model:

\[
Pr(y_{ijt} = 1 | x_{ijt}, u_{ijt-1}) = \Phi(\alpha'x_{ijt} + \tau_i + \delta_u u_{ijt-1} + \beta_{1i} u_{ijt-1} + \beta_{2i} d_{ij} + \beta_{3i} d_{ij} \times u_{ijt-1}),
\]

where the dependent variable \(y_{ijt}\) takes the value of one if an unemployed or employed worker was observed searching for a job at time \(t - 1\) and has found a job by the next calendar year and zero if he did not find a job within that calendar year. The vector \(x_{ijt}\) includes a set of individual controls for individual \(i\) living in state \(j\) at time \(t\). The terms \(\tau_i\) and \(\delta_u\) are state and year effects, and \(u_{ijt-1}\) is an unemployment dummy, which takes the value of one if the person was unemployed and the value of zero if the person was employed and searching for another job at time \(t - 1\). The parameter \(\beta_{1i}\) captures differences in year effects by employment status, and \(\beta_{2i}\) captures differences in state effects by employment status. The set of dummies, \(d_{ij}\), indicate coverage by the three unjust-dismissal exceptions, which take the value of one if the individual is observed living in a state \(j\) that has adopted the implied contract, public policy, or good faith exceptions at time \(t\) and zero otherwise. The vector \(\beta_3\) thus measures the direct impact of unjust-dismissal exceptions on employed workers. Finally, the vector \(\beta_4\) measures the differential impact of unjust-dismissal exceptions on the unemployed relative to employed workers.

Table 4 presents estimates from the probit of the direct impact of unjust-dismissal exceptions on employed workers and their relative impact on the unemployed, \(\beta_2\) and \(\beta_3\), as well as the impact of unemployment status on job finding probabilities, \(\delta\). The first four columns of table 4 present
Table 4
Job Finding Results

<table>
<thead>
<tr>
<th>Sample</th>
<th>Basic Controls</th>
<th>Basic + Main Effects, and Region × Time</th>
<th>Basic + Main Effects, and Region × Time and UI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td></td>
<td>(4)</td>
<td>(5)</td>
<td>(6)</td>
</tr>
<tr>
<td></td>
<td>(7)</td>
<td>(8)</td>
<td>(9)</td>
</tr>
<tr>
<td></td>
<td>(10)</td>
<td>(11)</td>
<td>(12)</td>
</tr>
<tr>
<td>Implied contract:</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect on employed</td>
<td>.086</td>
<td>.088</td>
<td>−.013</td>
</tr>
<tr>
<td></td>
<td>(.005)</td>
<td>(.009)</td>
<td>(.11)</td>
</tr>
<tr>
<td>Unemployment interaction</td>
<td>−.175</td>
<td>−.174</td>
<td>−.074</td>
</tr>
<tr>
<td></td>
<td>(.001)</td>
<td>(.009)</td>
<td>(.33)</td>
</tr>
<tr>
<td>Public policy:</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect on employed</td>
<td>.064</td>
<td>.046</td>
<td>−.033</td>
</tr>
<tr>
<td></td>
<td>(.005)</td>
<td>(.01)</td>
<td>(.011)</td>
</tr>
<tr>
<td>Unemployment interaction</td>
<td>−.091</td>
<td>−.033</td>
<td>−.032</td>
</tr>
<tr>
<td></td>
<td>(.001)</td>
<td>(.01)</td>
<td>(.001)</td>
</tr>
<tr>
<td>Good faith:</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect on employed</td>
<td>.008</td>
<td>−.053</td>
<td>−.001</td>
</tr>
<tr>
<td></td>
<td>(.006)</td>
<td>(.002)</td>
<td>(.009)</td>
</tr>
<tr>
<td>Unemployment interaction</td>
<td>−.073</td>
<td>−.046</td>
<td>−.122</td>
</tr>
<tr>
<td></td>
<td>(.001)</td>
<td>(.001)</td>
<td>(.048)</td>
</tr>
<tr>
<td>Unemployment main effect</td>
<td>.429</td>
<td>.409</td>
<td>.379</td>
</tr>
<tr>
<td></td>
<td>(.025)</td>
<td>(.028)</td>
<td>(.018)</td>
</tr>
</tbody>
</table>

Note.—The table reports marginal effects from regressions of re-employment probabilities on the interaction terms and main effects listed in the regressor column. The interaction terms capture the effects of the exceptions on unemployed relative to employed workers. All models include age, education, number of children, tenure, wage, nonwage income, local unemployment rate, a manufacturing dummy, a union dummy, a white-collar dummy, and dummies for race, sex, and marital status. Columns 5–8 include year effects, state effects, year-unemployed interactions, state-unemployed interactions, and region-time interactions. Columns 9–12 include, in addition, an unemployment benefit receipt dummy. Robust standard errors are reported in parentheses.
the results for the probit with basic controls.\textsuperscript{16} The coefficient $\beta_3$ in column 1 shows that the implied contract exception reduces the job finding probability of unemployed workers by 0.175 relative to employed workers. Columns 2 and 3 show smaller effects for unemployed workers relative to employed workers covered by the public policy and good faith doctrines, 0.091 and 0.073, respectively. Column 4 shows similar results when all three exceptions are included.\textsuperscript{17} These results also indicate that the job finding probability of unemployed workers is about 0.4 higher than for employed workers.

Columns 5–8 in table 4 show the results of the probits with the basic controls as well as all first-level and second-level main effects: time and state effects and time-unemployed and state-unemployed interactions. Time effects are included to capture the possibility that the introduction of exceptions may have coincided with other changes that were instead responsible for the low job finding rates. State effects are included because the exceptions may be capturing the low job finding rates of workers in those states that introduced exceptions for reasons unrelated to the exceptions. The time-unemployed and state-unemployed interactions are included to control for time-specific and state-specific factors affecting the unemployed but unrelated to the unjust-dismissal doctrines. In addition, these probits include interaction between time and region effects to control for time-varying regional shocks that may affect accession rates. Columns 5 and 6 in table 4 indicate smaller effects of the implied contract and public policy exceptions on the job finding probability of the unemployed relative to employed workers, after controlling for time, state, and time-unemployed, state-unemployed, and time-region interaction terms. Column 7, however, shows a larger effect of the good faith exception when all first-level and second-level terms are included, indicating that this exception reduces the job finding probability of unemployed workers by 0.122. This corresponds to a reduction of about 23\% in the unemployed’s job finding probability from 0.516 to 0.394. The results in column 8, including all exceptions, show that the exceptions together reduced the job finding probability of the unemployed by 0.161 relative to employed workers.

Columns 9–12 in table 4 present results for the probits including the basic controls, all first- and second-level main effects, and also an unemployment benefit receipt variable. The unemployment benefit variable

\textsuperscript{16} The controls include age, education, number of children, tenure, wage, non-wage income, local unemployment rate, dummies for race, sex, and marital status, a manufacturing dummy, a union dummy, and a white-collar dummy. The job-related variables (i.e., tenure, wage, union, sector, and occupation) for the unemployed refer to their previous job.

\textsuperscript{17} The estimates in cols. 1–4 are quite precise because these models omit main state and year effects.
is included because the lower job finding probability for the unemployed in states with exceptions may be capturing the higher propensity for the unemployed to claim benefits in these states. Columns 9–12 show very similar effects of the exceptions after controlling for unemployment benefit receipt.

The results in table 4 show a negative impact of dismissal costs on the job finding probabilities of unemployed relative to employed workers. Our results are robust to the inclusion of time and state effects, time-specific and state-specific effects on the unemployed, time-region interactions, and unemployment benefit receipt.18

Tables 5 and 6 present additional evidence consistent with the lemons story. Following Gibbons and Katz (1991), in table 5, we examine the impact of dismissal costs on nonunion and union workers. Since firms have less discretion firing workers covered by collective-bargaining agreements, the lemons effect generated by dismissal costs should be smaller for union workers. Columns 1–4 in table 5 show the results for the nonunion subsample, and columns 5–8 show results for the union subsample.19 Consistent with a lemons story, the estimates suggest that the implied contract and public policy doctrines had larger effects on nonunionized than unionized workers.

Table 6 presents results by reason for separation. It is widely believed that firms use temporary contracts as a way of avoiding dismissal costs and also as a way of screening workers (see, e.g., Autor 2001). One may thus expect for workers who have separated due to the end of one of these contracts to be stigmatized by unemployment in the same way dismissed workers are. At the same time, stigma effects should become less important for those under temporary contracts relative to those under permanent contracts as dismissal costs rise. As dismissal costs rise, firms will use more discretion in dismissing those under permanent contracts but not those under temporary contracts. Table 6 shows estimates of

18 To further probe the robustness of the results, we estimate probits that control for the maximum available welfare benefits (including AFDC benefits and food stamps) in a state at each point in time, since during this time period not only did unjust-dismissal legislation change differentially across states but also the generosity of welfare benefits. The results including welfare benefits are similar to those reported in table 4 and even slightly larger. In addition, we estimate models allowing for differential effects of the basic controls on unemployed and employed workers. The magnitude of the effects is somewhat smaller but remains large and significant. Finally, we estimate analogous linear probability and logit models with and without individual fixed effects. The resulting individual fixed-effects estimates are less precise but also suggest that the exceptions had large negative effects on the unemployed relative to employed workers.

19 There are too few observations in the union/good faith sample to be able to estimate the impact of the good faith doctrine on unionized workers.
Table 5  
Job Finding Results in Union and Nonunion Samples

<table>
<thead>
<tr>
<th>Exception and Regressor</th>
<th>Nonunion (1)</th>
<th>Nonunion (2)</th>
<th>Nonunion (3)</th>
<th>Nonunion (4)</th>
<th>Union (5)</th>
<th>Union (6)</th>
<th>Union (7)</th>
<th>Union (8)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Implied contract:</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect on employed</td>
<td>.089</td>
<td>.083</td>
<td>−.134</td>
<td>−.173</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(.011)</td>
<td>(.041)</td>
<td>(.018)</td>
<td>(.008)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unemployment interaction</td>
<td>−.195</td>
<td>−.193</td>
<td>−.127</td>
<td>−.105</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(0.004)</td>
<td>(.002)</td>
<td>(.01)</td>
<td>(.026)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Public policy:</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect on employed</td>
<td>.093</td>
<td>.057</td>
<td>.202</td>
<td>.201</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(0.03)</td>
<td>(.022)</td>
<td>(.151)</td>
<td>(.008)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unemployment interaction</td>
<td>−.11</td>
<td>−.041</td>
<td>−.094</td>
<td>−.057</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(0.003)</td>
<td>(.011)</td>
<td>(.036)</td>
<td>(.085)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Good faith:</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Effect on employed</td>
<td>−.024</td>
<td>−.076</td>
<td>−.076</td>
<td>−.076</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(0.025)</td>
<td>(0.035)</td>
<td>(0.035)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unemployment interaction</td>
<td>−.078</td>
<td>.053</td>
<td>.053</td>
<td>.107</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>(0.008)</td>
<td>(0.007)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Unemployment main effect</td>
<td>.427</td>
<td>.407</td>
<td>.437</td>
<td>.493</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(.03)</td>
<td>(.035)</td>
<td>(.014)</td>
<td>(.011)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N</td>
<td>4,244</td>
<td>4,244</td>
<td>4,244</td>
<td>4,244</td>
<td>864</td>
<td>864</td>
<td>864</td>
<td>864</td>
</tr>
</tbody>
</table>

Note.—The table reports marginal effects from regressions of re-employment probabilities on the interaction terms and main effects listed in the regressor column. The interaction terms capture the effects of the exceptions on unemployed relative to employed workers. The models also include age, education, number of children, tenure, hourly wage, nonwage income, local unemployment rate, a manufacturing dummy, a white-collar dummy, dummies for race, sex, and marital status, year effects, state effects, year-unemployed interactions, and state-unemployed interactions. The union/good faith sample (in col. 7) is too small to allow estimation of the effects. Robust standard errors are reported in parentheses.
models including interactions with a dismissal dummy. These results indicate higher job finding probabilities of dismissed relative to end-of-contract workers in employment-at-will states but smaller relative job finding probabilities for dismissed workers covered by the exceptions.

V. Conclusion

The matching model with asymmetric information presented in this article shows that firing costs are likely to generate hiring discrimination against the unemployed. This is true whether or not meeting rates are endogenous. Estimates using the NLSY indicate increased discrimination in hiring against the unemployed in the United States over the 1980s in those states that introduced exceptions to employment at will. These results are unchanged by including state and time effects, time-specific and state-specific effects on the unemployed, interactions between region and time effects, unemployment benefit receipt, and welfare benefits. Moreover, consistent with a lemons story, we find that the relative effect of the exceptions on the unemployed is generally smaller for unionized workers, who are subject to layoff-by-seniority rules, and for those who

---

Table 6
Job Finding Results by Reason for Separation

<table>
<thead>
<tr>
<th>Regressor</th>
<th>Basic Controls</th>
<th>Basic + Main Effects</th>
<th>Basic + Main Effects and UI</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
</tr>
<tr>
<td>Exception's effect on employed</td>
<td>-.001</td>
<td>-.397</td>
<td>-.392</td>
</tr>
<tr>
<td></td>
<td>(.009)</td>
<td>(.066)</td>
<td>(.055)</td>
</tr>
<tr>
<td>Exception × unemployment interaction</td>
<td>.041</td>
<td>.438</td>
<td>.443</td>
</tr>
<tr>
<td></td>
<td>(.009)</td>
<td>(.137)</td>
<td>(.122)</td>
</tr>
<tr>
<td>Exception × dismissal interaction</td>
<td>-.024</td>
<td>-.048</td>
<td>-.05</td>
</tr>
<tr>
<td></td>
<td>(.027)</td>
<td>(.021)</td>
<td>(.02)</td>
</tr>
<tr>
<td>Unemployment main effect</td>
<td>-.399</td>
<td>.325</td>
<td>.324</td>
</tr>
<tr>
<td></td>
<td>(.034)</td>
<td>(.184)</td>
<td>(.111)</td>
</tr>
<tr>
<td>Dismissal main effect</td>
<td>.487</td>
<td>.34</td>
<td>.335</td>
</tr>
<tr>
<td></td>
<td>(.002)</td>
<td>(.112)</td>
<td>(.124)</td>
</tr>
<tr>
<td>N</td>
<td>790</td>
<td>768</td>
<td>768</td>
</tr>
</tbody>
</table>

Note.—The table reports marginal effects from regressions of re-employment probabilities on the interaction terms and main effects listed in the regressor column. The models also include: age, education, number of children, tenure, hourly wage, nonwage income, local unemployment rate, a manufacturing dummy, a white-collar dummy, a union dummy, dummies for race, sex, and marital status, year effects, state effects, year-unemployed interactions, and state-unemployed interactions. Robust standard errors are reported in parentheses.
lost their jobs due to an end of contract and, thus, are not subject to dismissal costs.

While our empirical analysis used U.S. data, the results also have implications for European labor markets. Since European countries have high dismissal costs as compared to North American countries, European firms should discriminate even more against unemployed job seekers. These predictions are consistent with the findings in Bertola and Rogerson (1997) and Boeri (1999), which show much lower flows into and out of unemployment but similar job-to-job flows in the two continents. Our article suggests that employment protection legislation, together with information asymmetries, probably plays an important role in explaining these differences.

Appendix

Properties of Equation (4)

The right-hand side of equation (4) is increasing in the dismissal threshold \( m_\eta \), so that equation (4) determines \( m_\eta \) uniquely. Totally differentiating equation (4) with respect to \( \eta \), \( \bar{w} \), and \( F \), we get

\[
\frac{dm_\eta}{d\eta} = -1 - \gamma \frac{d\bar{f}(\eta)}{(1 - \phi) d\eta},
\]

\[
\frac{dm_\eta}{d\bar{w}} = 1 - \frac{\gamma}{(1 - \phi)} \frac{d\bar{f}(\eta)}{d\bar{w}},
\]

\[
\frac{dm_\eta}{dF} = \frac{(r + \gamma + \alpha)}{(1 - \phi)} - \frac{\gamma}{(1 - \phi)} \frac{d\bar{f}(\eta)}{dF},
\]

and differentiating equation (5) with respect to \( \eta \), \( \bar{w} \), and \( F \) and using the envelope theorem, we obtain

\[
\frac{d\bar{f}(\eta)}{d\eta} = \frac{(1 - \phi)}{(r + \gamma)} (G(\bar{m}) - G(\bar{m})) + \frac{(1 - \phi)}{(r + \gamma + \alpha)} (G(\bar{m}) - G(m)) > 0,
\]

\[
\frac{d\bar{f}(\eta)}{d\bar{w}} = -\frac{d\bar{f}(\eta)}{d\eta} < 0,
\]

\[
\frac{d\bar{f}(\eta)}{dF} = \frac{-G(m(\eta))}{1 - \frac{\gamma}{(r + \gamma)} (G(\bar{m}) - G(\bar{m})) + \frac{\gamma}{(r + \gamma + \alpha)} (G(\bar{m}) - G(m))} < 0.
\]

Substituting these into the above equations, we obtain that \( dm_\eta/d\eta < 0 \), \( dm_\eta/d\bar{w} > 0 \), and \( dm_\eta/dF < 0 \).

Furthermore, we can see that the dismissal threshold of good workers
Thus, the effects of the second and fourth terms in the brackets with respect to are zero. Given that \( dm_r / d\eta < 0 \) and, thus, \( m_r(\eta_{x}) < m_r(\eta_{d}) \), then \( 0 > df(\eta_{x})/dF > df(\eta_{d})/dF \), and we get that

\[
0 > dm_r(\eta_{x})/dF > dm_r(\eta_{d})/dF.
\]

Similarly, given that \( m_r(\eta_{d}) < m_r(\eta_{x}) \), then \( 0 > df(\eta_{d})/d\bar{w} > df(\eta_{x})/d\bar{w} \), and

\[
\frac{dm_r(\eta_{d})}{d\bar{w}} > \frac{dm_r(\eta_{x})}{d\bar{w}} > 0,
\]

which proves that there is a greater response of \( m_r(\eta_{x}) \) than of \( m_r(\eta_{d}) \) to changes in \( F \) and \( \bar{w} \).

**Properties of EB**

We can write \( \Pi_s \) as a function of \( \bar{z}_s \) and the exogenous parameters of the model. Then \( f(\bar{m}, \eta_{x}) \) and \( f(\bar{m}, \eta_{d}) \) can be computed using (3) and substituting \( f(\theta) \), which only depends on \( m(\eta) \) and on exogenous parameters. Given that \( m(\eta) \) is a sole function of such parameters, \( \Pi_s \) can be written as a function of \( \bar{z}_s \) and exogenous parameters,

\[
\bar{z}_s = \Pi_s = z_r f(\bar{m}, \eta_{x}) + (1 - z_r) f(\bar{m}, \eta_{d})
\]

\[
= z_r \left[ \frac{(1 - \varphi)(\bar{m} + \eta_r - \bar{\omega})}{r + \gamma + \alpha} \right]
+ \gamma \left[ \int_{\eta_d}^{\eta_x} \frac{(1 - \varphi)(m + \eta_r - \bar{\omega})}{r + \gamma} g(m) dm + \frac{(1 - \varphi)(m + \eta_r - \bar{\omega})}{r + \gamma + \alpha} \left( G(m) - G(\bar{m}) \right) \right]
+ (1 - z_r) \left[ \frac{(1 - \varphi)(\bar{m} + \eta_r - \bar{\omega})}{r + \gamma + \alpha} \right]
+ \gamma \left[ \int_{\eta_r}^{\eta_d} \frac{(1 - \varphi)(m + \eta_r - \bar{\omega})}{r + \gamma} g(m) dm + \frac{(1 - \varphi)(m + \eta_r - \bar{\omega})}{r + \gamma + \alpha} \left( G(m) - G(\bar{m}) \right) \right]
+ (1 - z_r) \left[ \frac{(1 - \varphi)(\bar{m} + \eta_r - \bar{\omega})}{r + \gamma + \alpha} \right]
+ \gamma \left[ \int_{\eta_r}^{\eta_x} \frac{(1 - \varphi)(m + \eta_r - \bar{\omega})}{r + \gamma} g(m) dm + \frac{(1 - \varphi)(m + \eta_r - \bar{\omega})}{r + \gamma + \alpha} \left( G(m) - G(\bar{m}) \right) \right]

(A1)

Furthermore, \( \partial \Pi_s / \partial \bar{z}_s = f(\bar{m}, \eta_{x}) - f(\bar{m}, \eta_{d}) > 0 \). Therefore, there exists a unique value of \( \bar{z}_s \) such that the condition \( \Pi_s = 0 \) is satisfied. This defines a horizontal line PP, which delimits the plane between a region where \( \Pi_s > 0 \), in which case \( p_r = 1 \), and a region where \( \Pi_s < 0 \), in which case \( p_r = 0 \). This establishes the shape of the EB locus.

Next, totally differentiating (A1), we obtain that the derivatives of the second and fourth terms in the brackets with respect to \( m(\eta) \) are zero. Thus, the effects of \( F \) and \( \bar{w} \) on \( \bar{z}_s \) reduce to the direct effects of these
parameters on profits,

\[
\frac{dz_a}{dF} = \frac{c}{(r + \gamma + \alpha)(J(\tilde{m}, \eta_{\tilde{l}}) - J(\tilde{m}, \eta_l))} + \left(1 - \frac{z_a}{D_H} + \frac{(1 - z_a)}{D_L}\right) > 0,
\]

and

\[
\frac{dz_a}{d\tilde{w}} = \frac{(1 - \varphi)}{(r + \gamma + \alpha)(J(\tilde{m}, \eta_{\tilde{l}}) - J(\tilde{m}, \eta_l))} > 0,
\]

where

\[
D_H = \left[1 - \frac{\gamma}{(r + \gamma)}[G(\tilde{m}) - G(\tilde{m})] - \frac{\gamma}{(r + \gamma + \alpha)}[G(\tilde{m}) - G(m(\eta_{\tilde{l}}))]ight]
\]

and

\[
D_L = \left[1 - \frac{\gamma}{(r + \gamma)}[G(\tilde{m}) - G(\tilde{m})] - \frac{\gamma}{(r + \gamma + \alpha)}[G(\tilde{m}) - G(m(\eta_{\tilde{l}}))].
\]

Proof That S-S Is Downward Sloping

Differentiating equation (9) with respect to \( p_a \) shows that the sign of the slope is equal to the sign of the following expression:

\[
\frac{dz_a}{dp_a} \propto \gamma z(1 - z) \left(1 - \frac{1}{G_l} - \frac{1}{G_H}\right),
\]

which is negative since \( G_H < G_L \). Q.E.D.

Proof of Proposition 1

Differentiating equation (9), while holding \( p_a \) constant, the direction of the move of the S-S locus in response to an increase in \( F \) is of the same sign as

\[
\frac{dz_a}{dF} \propto -\gamma \left[ \frac{g_l}{(G_l)^2} \frac{dm(\eta_l)}{dF} - \frac{g_H}{(G_H)^2} \frac{dm(\eta_{\tilde{l}})}{dF} \right] - \frac{\alpha p_a}{G_H G_l} \left[ \frac{g_l}{G_l} \frac{dm(\eta_l)}{dF} - \frac{g_H}{G_H} \frac{dm(\eta_{\tilde{l}})}{dF} \right].
\]

We know from the properties of equation (4) that \( 0 > \frac{dm(\eta_l)}{dF} > \frac{dm(\eta_{\tilde{l}})}{dF} \). Thus, given that \( G_l > G_H \) and the nonincreasing hazard assumption, \( dz_a/dF \) is clearly negative. Q.E.D.
Proof That If $F$ Is Low Enough, $p_r = 1$

Assume that $\bar{m} + \eta_H > \bar{w}$. This means that it is at least profitable for firms to employ good workers in the best possible state. At $F = 0$, one has $J(\bar{m}, \eta_H) > -\bar{F} = 0$ and $J(\bar{m}, \eta_1) \geq -\bar{F} = 0$, implying that $\Pi_r > 0$ for all $z_r$. By continuity, this property holds in the neighborhood of $F = 0$. Q.E.D.

Proof of Theorem 1

Equation (13) is equivalent to

$$C = \lambda_r \Pi_r + \lambda_u p_r \Pi_u. \tag{A2}$$

The equations of Section II as well as (11) and (12) determine, in a reduced form, the right-hand side of (A2) as a function of $a$. Inspection of the relevant equations reveals that this function is continuous. Call it $H(a)$. Next, note that as $a$ goes from any $a_0$ to infinity, the $\bar{x}$s are bounded from above by $q(a_z)$, while (3) implies that $\Pi_r$ and $\Pi_u$ go to zero. Consequently,

$$\lim_{a \to \infty} H(a) = 0.$$

Finally, if $H(0) \leq C$, then there exists an equilibrium such that $a = 0, u = 1$, that is, where no hiring is profitable. If not, then $H(0) > C$, in which case, by continuity, there exists an $a$ such that $H(a) = C$ and $H(\cdot)$ is locally decreasing around $a$. Q.E.D.

References


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Note that $J(m, \eta)$ has a discontinuity as $m$ goes through $\bar{m}$. However, what intervenes in the determination of the right-hand side of (A2) is $J(m, \eta)$, which is clearly continuous in $a$. 