

High Discounts and High Unemployment *

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Abstract

In recessions, the stock market falls more than in proportion to corporate profit. The discount rate implicit in the stock market rises. All types of investment fall, including employers' investment in job creation. According to the leading view of unemployment—the Diamond-Mortensen-Pissarides model—when the incentive for job creation falls, the labor market slackens and unemployment rises. Employers recover their investments in job creation by collecting a share of the surplus from the employment relationship. The value of that flow falls when the discount rate rises. Thus high discount rates imply high unemployment. This paper does not explain why the discount rate rises so much in recessions. Rather, it shows that the rise in unemployment makes perfect economic sense in an economy where the stock market falls substantially in recessions because the discount rises.

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The search-and-matching paradigm has come to dominate theories of movements of unemployment, because it has more to say about the phenomenon than merely reducing unemployment to the difference between labor supply and labor demand. The ideas of Diamond, Mortensen, and Pissarides promise a deeper understanding of fluctuations in unemployment, most recently following the worldwide financial crisis that began in late 2008. But connecting the crisis to high unemployment according to the principles of the DMP model has proven a challenge.

In a nutshell, the DMP model relates unemployment to job-creation incentives. When the payoff to an employer from taking on new workers declines, employers put fewer resources into recruiting new workers. Unemployment then rises and new workers become easier to find. Hiring returns to its normal level, so unemployment stabilizes at a higher level and remains there until job-creation incentives return to normal. This mechanism rests on completely solid ground.

The question about the model that is unresolved today, 20 years after the publication of the canon of the model, Mortensen and Pissarides (1994), is: What force depresses the payoff to job creation in recessions? In that paper, and in hundreds of successor papers, the force is a drop in productivity. But that characterization runs into two problems: First, unemployment did not track the movements of productivity in the last three recessions in the United States. Second, as Shimer (2005) showed, the model, with realistic parameter values, implies tiny movements in unemployment in response to large changes in productivity. This paper considers a different driving force, the discount rate employers apply to the stream of benefits they receive from a new hire.

A simple model lays out the issues in this paper. The economy follows a Markov process between a normal state, numbered $i = 1$, and a depressed state, numbered $i = 2$. I pick parameter values to approximate the U.S. labor market. The probability of exiting the normal state is $\pi_1 = 0.0083$ per month and the probability of exiting the depressed state is $\pi_2 = 0.017$ per month. The expected duration of a spell in the normal state is 10 years and the expected duration in the depressed state is 5 years. A worker has productivity 1 and receives a wage $w = 0.985$. Workers separate from their jobs with monthly hazard $s = 0.035$. Agents discount future profit $1 - w$ at the rate r_i , with $r_1 = 0.0083$ (10 percent per year) and $r_2 = 0.042$ (50 percent per year). The value of a worker to a firm is

$$J_1 = 1 - w + \frac{1 - s}{1 + r_1} [(1 - \pi_1)J_1 + \pi_1 J_2] \quad (1)$$

and similarly for J_2 . The solution is $J_1 = 0.32$ and $J_2 = 0.22$.

The labor market operates according to the search-and-matching principles of DMP. The matching function is Cobb-Douglas with equal elasticities for vacancies and unemployment. The monthly cost of maintaining a vacancy is $c = 0.43$. The market is in equilibrium when the cost of recruiting a worker equals the value of the worker:

$$cT_1 = \frac{1}{1 + r_1} [(1 - \pi_1)J_1 + \pi_1 J_2] \quad (2)$$

and similarly for $i = 2$. The expected duration of a vacancy is T_i months ($T_1 = 0.73$ months and $T_2 = 0.49$ months). The job-finding rate is $f_i = \mu^2 T_i$, where μ is the efficiency parameter of the matching function. The stationary unemployment rate is

$$u_i = \frac{s}{s + f_i}, \quad (3)$$

with $u_1 = 6.1$ percent and $u_2 = 8.9$ percent.

Unemployment rises in the depressed state because of the higher discount rate. This paper is about the depressing effect in the labor market of higher discounts. Two major research topics arise. First, I demonstrate that Nash bargaining cannot determine the wage. Not only must the wage be less responsive to the tightness of the labor market than it would be with Nash bargaining—a point well understood since Shimer (2005)—but the wage must be a simple markdown from productivity, as in the model above. This finding is new. The paper derives the markdown model from the large body of wage-determination theory in the DMP tradition and presents evidence supporting this specialization of the more general theory. The markdown property finds support in an important new paper, Chodorow-Reich and Karabarbounis (2013), on the time-series behavior of the opportunity cost of labor to the household.

Second, I demonstrate that the increase in the discount rate needed to generate a realistic increase in unemployment in a depressed period is substantial, far in excess of any increase in real interest rates. Thus the paper needs to document high discount rates in depressed times.

The causal chain I have in mind is that some event creates a financial crisis, in which risk premiums rise so discount rates rise, asset values fall, and all types of investment decline. In particular, the value that employers attribute to a new hire declines on account of the higher discount rate. Investment in hiring falls and unemployment rises. Of course, a crisis results in *lower* discount rates for safe flows—the yield on 5-year U.S. Treasury notes fell essentially to

zero soon after the crisis of late 2008. The logic pursued here is that the flow of benefits from a newly hired worker has financial risk comparable to corporate earnings, so the dramatic widening of the equity premium that occurred in the crisis implied higher discounting of benefit flows from workers at the same time that safe flows from Treasuries received lower discounting. In the crisis, investors tried to shift toward safe returns, resulting in lower equity prices from higher discount rates and higher Treasury prices from lower discounts. In other words, the driving force for high unemployment is a substantial widening of the risk premium for the future stream of contributions a new hire makes to an employer.

The appendix discusses some of the large number of earlier contributions to the DMP and finance literatures relevant to the ideas in this paper. The idea that the discount rate affects unemployment is not new. Rather, the paper's contribution is to connect the labor market to the finance literature on the volatility of discount rates in the stock market and to identify parameters of wage determination that square with the high response of unemployment to discount fluctuations and the low response of unemployment to productivity fluctuations.

1 The Job Value and the Stock Market

The job value J is the present value, using the appropriate discount rate, of the flow benefit that an employer gains from an added worker, measured as of the time the worker begins the job. The value of a stock-market price index is the present value, again using an appropriate discount rate, of the dividends that the holder of the index's portfolio receives in the future. On the hypothesis that the flow benefit of a worker and dividends have similar determinants rooted in the success of employers and that the discount rates reflect similar risks of the two future flows, the two present values ought to move together, under the view of the labor market developed in this paper. At times, notably in the past 20 years, this proposition has been spectacularly true.

1.1 The job value and equilibrium in the labor market

The incentive for a firm to recruit a new worker is the present value of the difference between the marginal benefit that the worker will bring to the firm and the compensation the worker will receive. In equilibrium, with free entry to job creation, that present value will equal the expected cost of recruitment. The cost depends on conditions in the labor market, measured by the number of job openings or vacancies, V , and the flow of hiring, H . A good

approximation, supported by extensive research on random search and matching, is that the cost of recruiting a worker is

$$\kappa + cx \frac{V}{H}. \quad (4)$$

Here x is labor productivity. The vacancy/hiring ratio $T = V/H$ is the expected time to fill a vacancy, so the parameter c is the per-period cost of holding a vacancy open, stated in labor units. The equilibrium condition is

$$\kappa + cxT = \frac{1}{1+r} \bar{J}. \quad (5)$$

\bar{J} is the present value of the new worker to the employer. I let $J = \bar{J} - (1+r)\kappa$, the net present value of the worker to the employer, so the equilibrium condition becomes

$$cxT = \frac{1}{1+r} J. \quad (6)$$

The DMP literature invariably uses the vacancy/unemployment ratio $\theta = V/U$ as the measure of tightness. Under the assumption of a Cobb-Douglas matching function with equal elasticities for unemployment and vacancies (hiring flow = $\mu\sqrt{UV}$), the relation between the two tightness measures is

$$\theta = \mu^2 T^2. \quad (7)$$

1.2 Pre- and post-contract costs

The DMP model rests on the equilibrium condition that the employer anticipates a net benefit of zero from starting the process of job creation. An employer considering recruiting a new worker expects that the costs sunk at the time of hiring will be offset by the excess of the worker's contribution over the wage during the ensuing employment relationship. The model makes a distinction between costs that the employer incurs to recruit job candidates and costs incurred to train and equip a worker. In the case that an employer incurs training costs, say K , immediately upon hiring a new worker, and then anticipates a present value \tilde{J} from the future flow benefit—the difference $x - w$ between productivity and the wage—the equilibrium condition would be

$$\tilde{J} - K - (1+r)cT = 0. \quad (8)$$

In this case, the job value considered here would be the net, pre-training value, $J = \tilde{J} - K$. The job value \tilde{J} rises by the amount K when the training cost is sunk.

Notice that training costs have a role similar to that of the constant element of recruiting, κ . The definition of J used here isolates a version of the job value that is easy to observe and moves the hard-to-measure elements to the right-hand side. Thus training and other startup costs and the fixed component of recruiting cost are deductions from the present value of $x - w$ in forming J as it is defined here.

Costs not yet incurred at the time that the worker and employer make a wage bargain are a factor in that bargain. The employer cannot avoid the pre-contract cost of recruiting, whereas the post-contract training and other startup costs are offset by a lower wage and so fall mainly on the worker under a standard calibration of the bargaining problem.

1.3 Measuring the job value

The labor market reveals the job value from the condition that the value equals the cost of attracting an applicant, which is the per-period vacancy cost times the duration of the typical vacancy, brought forward one period: $J = (1 + r)cxT$. Later in the paper, I discuss the challenges to the alternative way of evaluating J as the present value of a flow $x - w$ measured from independent data on x and w .

Data from Silva and Toledo (2009) show that the daily cost of maintaining a vacancy is 0.43 days of pay, so $c = \$66$ per day for the average U.S. employee in January 2011. I use this value to calculate J , but the main results of the paper do not depend on knowing the value of c .

The BLS's Job Openings and Labor Turnover Survey (JOLTS) reports the number of vacancies and the hiring rate. The average duration of a vacancy is the ratio of the two. Figure 1 shows the result of the calculation for the total private economy starting in December 2000, at the outset of JOLTS, through the beginning of 2013. The average job value over the period was \$1080 per newly hired worker. The value started at \$1506, dropped sharply in the 2001 recession and even more sharply and deeply in the recession that began in late 2007 and intensified after the financial crisis in September 2008. The job value reached a maximum of \$1,467 in December 2007 and a minimum of \$769 in July 2009. Plainly the incentive to create jobs fell substantially over that interval. Hall and Schulhofer-Wohl (2013) compare the hiring flows from JOLTS to the total flow into new jobs from unemployment, those out of the labor force, and job-changers. The level of the flows is higher in the CPS

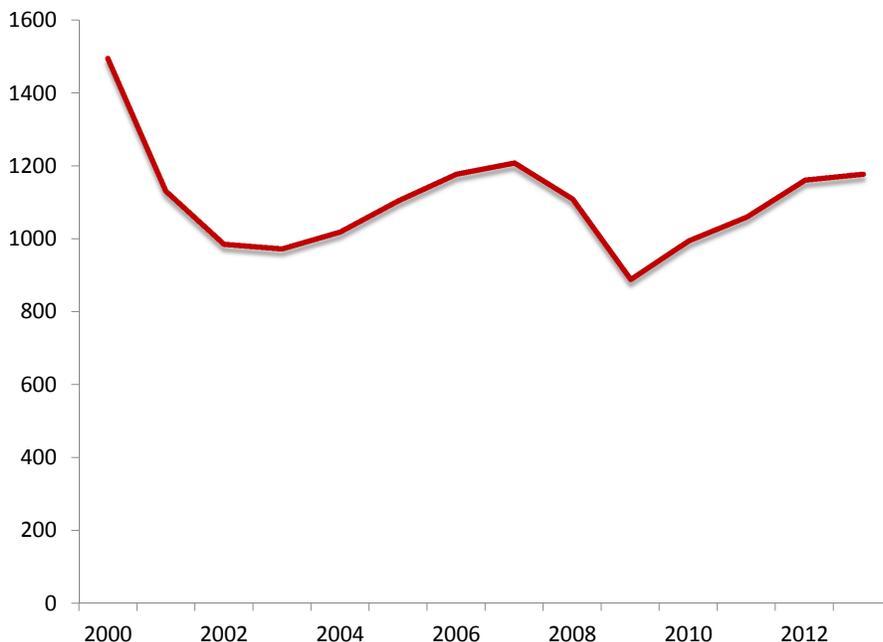


Figure 1: Aggregate Job Value, 2001 through 2013

data and the decline in the recession was somewhat larger as well. None of the results in this paper would be affected by the use of the CPS hiring flow in place of the JOLTS flow.

Figure 2 shows similar calculations for the industries reported in JOLTS. Average job values are lowest in construction, which fits with the short duration of jobs in that sector. The highest values are in government and health. Large declines in job values occurred in every industry after the crisis, including health, the only industry that did not suffer declines in employment during the recession. The version of the DMP model developed here explains the common movements of job values across industries, including those that have employment growth, as the common response to the increase in the discount rate.

Lack of reliable data on hiring flows prevents the direct calculation of job values prior to 2001. Data are available for the vacancy/unemployment ratio. I will discuss this source shortly. From it, an approximation to the duration of vacancies measure is available as

$$T = \frac{\sqrt{\theta}}{\mu}, \quad (9)$$

using the years 2001 through 2007 to measure matching efficiency μ (efficiency dropped sharply beginning in 2008). Figure 3 shows the job-value proxy. It is a completely reliable cyclical indicator, negatively correlated with unemployment.

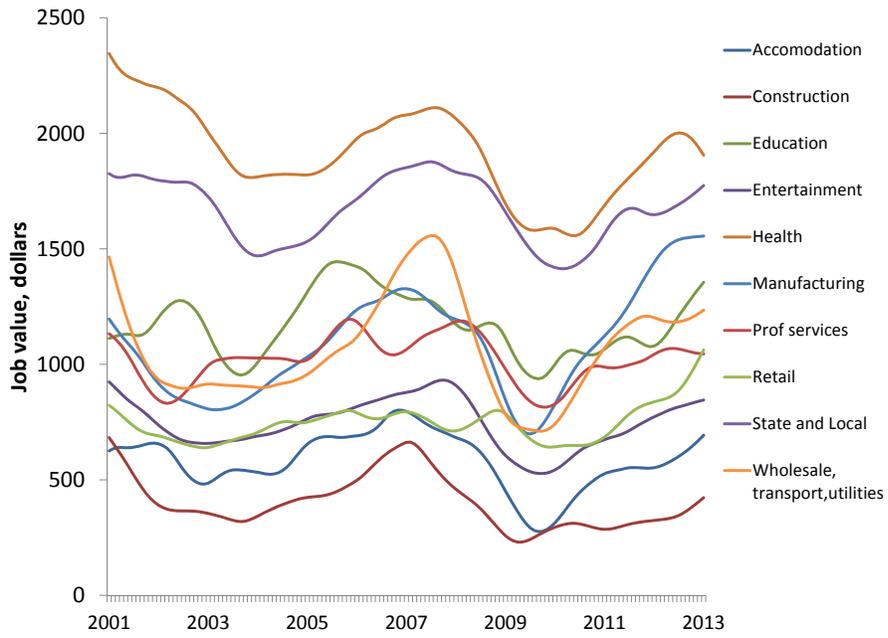


Figure 2: Job Values by Industry, 2001 through 2013

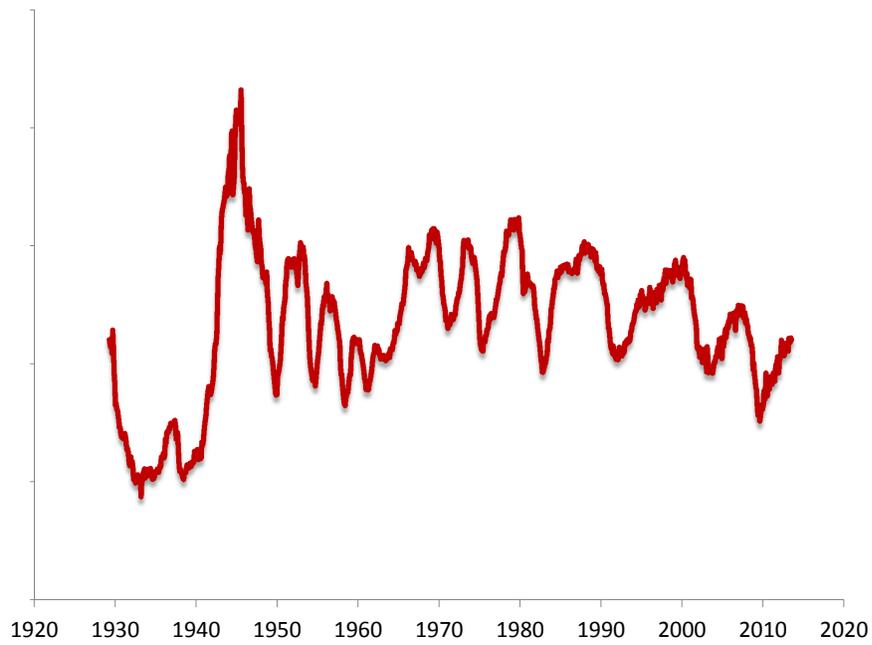


Figure 3: Proxy for the Job Value, 1929 through 2013

1.4 The relation between the job value and the stock market

Kuehn, Petrosky-Nadeau and Zhang (2013) show that, in a model without capital, the return to holding a firm's stock is the same as the return to hiring a worker. In levels, the same proposition is that the value of the firm in the stock market is the value of what it owns. Under a policy of paying out earnings as dividends, rather than holding securities or borrowing, the firm without capital owns only one asset, its relationships with its workers. The stock market reveals the job value of workers (the amount cT) plus any other costs the firm incurred with the expectation that they would be earned back from the future difference between productivity and wage, $x - w$. Of course, in reality firms also own plant and equipment. One could imagine trying to recover the job value by subtracting the value of plant and equipment and other assets from the total stock-market value. Hall (2001) suggests that the results would not make sense. In some eras, the stock-market value falls far short of the value of plant and equipment alone, while in others, the value is far above that benchmark, much further than any reasonable job value could account for. The appendix discusses Merz and Yashiv's (2007) work relating plant, equipment, and employment values to the stock market.

1.5 Comparison of the job value to the value of the stock market

Figure 4 shows the job value calculated earlier, together with the S&P 500 index of the broad stock market, deflated by the Consumer Price Index scaled to have the same mean as the job value. The S&P 500 includes about 80 percent of the value of publicly traded U.S. corporations but omits the substantial value of privately held corporations. The similarity of the job value and the stock-market value is remarkable. The figure strongly confirms the hypothesis that similar forces govern the market values of claims on jobs and claims on corporations.

Figure 5 shows the the relation between the job-value proxy and the detrended S&P stock-market index (now the S&P500) over a much longer period. I believe that the S&P is the only broad index of the stock market available as early as 1929. The figure confirms the tight relation between the job value and the stock market in the 1990s and later, and also reveals other episodes of conspicuous co-movement. On the other hand, the figure is clear that slow-moving influences differ between the two series in some periods. During the time

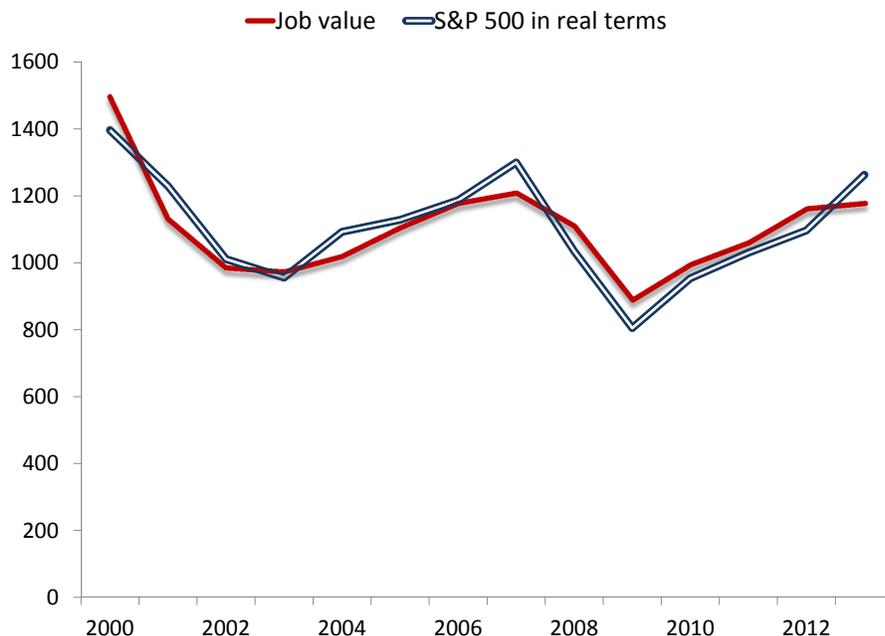


Figure 4: Job Value from JOLTS and S&P Stock-Market Index, 2001 through 2013

when the stock market had an unusually low value by almost any measure, from the mid-70s through 1991, the two series do not move together nearly as much.

Figure 6 shows the co-movement of the job value and the stock market at business-cycle frequencies. It compares the two-year log-differences of the job-value proxy and the S&P index. It supports the conclusion that the two variables share a common cyclical determinant.

The similarity of the movements of the two variables indicates that the job value—and therefore the unemployment rate—shares its determinants with the stock market. This finding supports the hypothesis that rises in discount rates arising from common sources, such as financial crises, induce increases in unemployment. In both the labor market and the stock market, the value arises from the application of discount rates to expected future flow of value. The next step in this investigation is to consider the discount rates and the value flows subject to discount separately.

2 The Discount Rate in the DMP Model

2.1 Equilibrium with Nash wage bargain

The following analysis comes directly from Pissarides (2000), chapter 1, which contains a full exposition of the DMP model with Nash bargaining, including a discussion of the role

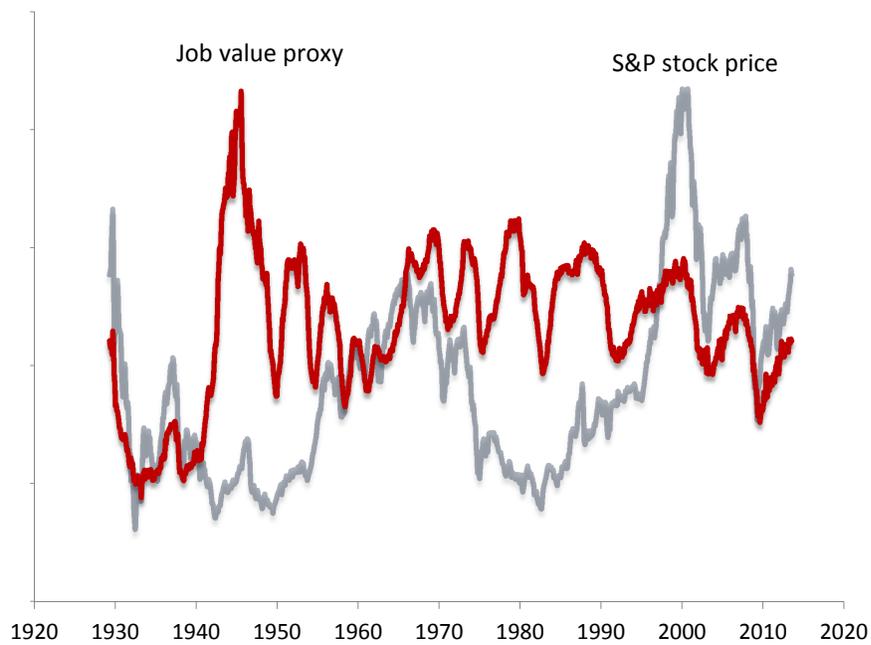


Figure 5: Job-Value Proxy and the S&P Stock-Market Index

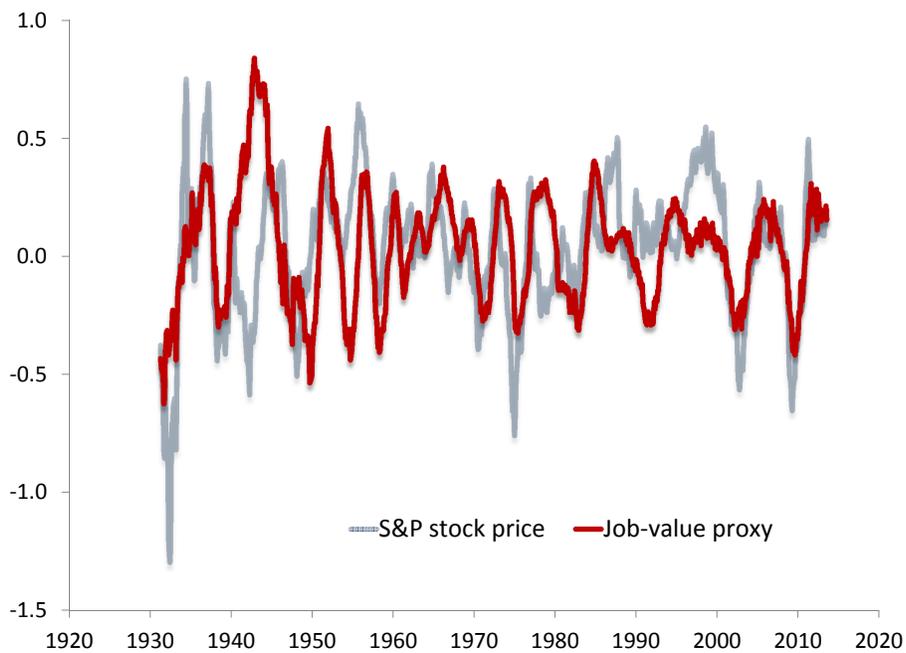


Figure 6: Two-Year Log-Differences of the Job Value and the S&P Stock-Market Price Index

of the discount rate. He derives an informative expression for the Nash-bargained wage in the canonical DMP model:

$$w = (1 - \beta)z + \beta x(1 + c\theta). \quad (10)$$

Here β is the bargaining power of the jobseeker, z is the flow value of unemployment as a fraction of the normal value of x , and θ is the vacancy/unemployment ratio. The presence of θ arises from the benefit that a jobseeker gains when bargaining in a tighter market with more vacancies per jobseeker. With separation rate s , the stationary job value is

$$J = \frac{1 + r}{r + s}(x - w). \quad (11)$$

Equation (6), equation (10), and equation (11) combine to form the equilibrium condition:

$$\frac{(1 - \beta)(x - z) - \beta x c \theta}{r + s} = c x \frac{\sqrt{\theta}}{\mu}. \quad (12)$$

Provided that the labor market meets the basic condition for viability, that productivity x exceed the flow value of unemployment, z , the equilibrium condition has a unique root θ .

To calibrate the model at a monthly frequency, I take (as in the introduction) $r = 0.10/12$, $s = 0.035$ (from JOLTS), and $c/x = 0.43$ (from Silva and Toledo (2009)). The average vacancy/unemployment ratio starting in 1948 is $\theta = 0.44$. From the average unemployment rate since 1948 of 0.058, I find matching efficiency $\mu = 0.86$, job-finding rate 0.57 per month, job-filling rate 1.21 hires per vacancy per month, and vacancy duration $T = 0.83$ months. I take the bargaining weight to be $\beta = 0.5$, a reasonable value that also is a limiting case of the alternating-offer bargaining model I will mention shortly. Finally, I choose the flow value of unemployment to satisfy the equilibrium condition at the calibration point: $z = 0.78$, somewhat higher than the value of 0.71 in Hall and Milgrom (2008).

Most of the research on unemployment volatility has focused on productivity as the driving force. In the model as calibrated, a decrease in productivity of one percent (a fairly large shock by historical standards) raises unemployment from 5.80 percent to 5.91 percent. Thus the calibration replicates the finding of Shimer (2005)—the reasonably calibrated DMP model with equal bargaining weights requires huge fluctuations in productivity to generate the large observed movements in unemployment.

The same point carries over to fluctuations in the discount rate. Raising the annual rate from 10 percent to 20 percent (also a fairly large shock) raises unemployment from 5.80

percent to 5.88 percent. The canonical DMP model is equally unsuccessful in generating observed movements in unemployment from realistic movements in the discount rate.

The reason that unemployment is so stable in the canonical DMP model is that wages move to offset changes in driving forces. A powerful equilibrating mechanism operates through θ in equation (10). A decline in productivity or a rise in the discount rate causes an incipient decline in θ . If the labor market actually softened, workers would be in a weaker bargaining position and the Nash-bargained wage would fall. The elasticity of the job value J with respect to the wage is -68 at the calibration point. A small drop in the wage offsets almost all of the effect of the decline in productivity or rise in the discount rate.

2.2 Generalizing the DMP model by attenuating the tightness effect on the wage

The high sensitivity to labor-market tightness is a property of the symmetric Nash bargain, because it gives substantial weight to the worker's threat to disclaim the wage negotiation and continue searching. The value of searching is higher when the market is tight. A bargaining protocol that gives less weight to the threat to look elsewhere will generate higher and more realistic sensitivity to driving forces, including the discount rate and productivity.

One way to capture this idea in a simple way is to introduce a parameter ψ that controls the role of tightness in the wage bargain. I relabel the Nash wage from equation (10) as w^N . I define a wage without the tightness effect as

$$w^x = (1 - \beta)z + \beta x(1 + c\bar{\theta}), \quad (13)$$

where $\bar{\theta}$ is a constant equal to the value of θ at the calibration point (so that the same calibration continues to apply). Then the bargained wage is the weighted average of w^N and w^x :

$$w = \psi w^N + (1 - \psi)w^x. \quad (14)$$

With $\psi = 0$ —full insulation of the wage from the influence of tightness—the increase in the discount rate to 20 percent annually raises unemployment from 5.8 to 6.8 percent and the decline of one percent in productivity raises unemployment to 7.8 percent. This view of wage bargaining makes both driving forces capable of generating fluctuations of realistic size in unemployment.

The bargaining protocol that results in $\psi < 1$ is plausible but not founded in any formal bargaining model. Hall and Milgrom (2008) provide such a model. That paper points

out that a jobseeker’s threat to break off wage bargaining and continue to search is not credible, because the employer—in the environment described in the basic DMP model with homogeneous workers—always has an interest in making a wage offer that beats the jobseeker’s option of breaking off bargaining. Similarly, the jobseeker always has an interest in making an offer to the employer that beats the employer’s option of breaking off bargaining and forgoing any profit from the employment opportunity. Neither party, acting rationally, would block the employment bargain when doing so throws away the joint value. The alternating-offer bargaining model captures this idea. Our paper shows that the resulting bargain remains sensitive to productivity but loses most of its sensitivity to labor-market tightness, because that sensitivity arises in the Nash setup only because of the unrealistic role of the non-credible threat to break off bargaining and continue to search.

Hall and Milgrom’s credible-bargaining model has a parameter with the same effect as ψ , called δ . It is the per-period probability that some external event will destroy the job opportunity and send the jobseeker back into the unemployment pool. If that probability is zero, the model delivers maximal insulation from tightness, whereas if it is one, the alternating-offer model is the same as the symmetric Nash bargaining model. Thus δ defines one dimension of the parameter space in the same way as ψ . It is straightforward to generalize the credible-bargaining model to link the flow value of unemployment, z , to productivity via the parameter α —see the appendix.

2.3 Generalizing the DMP model by linking z to productivity

Another potential variation in the DMP model will be important in the following discussion. Tightness in the model depends on the gap between productivity x and the flow value of unemployment, z . This gap indexes the benefit of employment over unemployment—see equation (12). An implicit assumption in the amplification effect just derived—and in almost all of the DMP literature—is that a force that causes productivity to fall has no effect on z . A more general view would have z change when x changes:

$$z = [(1 - \alpha)x + \alpha\bar{x}]\bar{z}. \tag{15}$$

Here \bar{x} is the level of productivity at the calibration point and \bar{z} is the level of z at that point. The standard view has $\alpha = 1$. Low values of α result in small sensitivity of tightness and unemployment to changes in productivity, with no difference in the response to changes in the discount rate. With $\alpha = 0$ (z moving in proportion to x), changes in productivity have

no effect on tightness. In this case, the wage is a markdown on productivity, as mentioned in the introduction.

Notice the key distinction between a *sticky* wage—one less responsive to all of its determinants—and a *tightness-insulated* wage. The latter responds substantially to productivity while attenuating the Nash bargain’s linkage of wages to the ease of finding jobs. Something like the tightness-insulated wage is needed to rationalize the strong relation between the discount rate and the unemployment rate discussed in this paper. With $\alpha = 0$, tightness-insulation is maximal.

The two parameters ψ and α define a space within which any combination of positive effects of productivity on unemployment and negative effects of the discount rate on unemployment can occur. In this paper, I make a case for low values of ψ and α . The case is easy to explain: Given values ψ and α , and time series for the observed values of productivity x_t and unemployment u_t , the model implies values of the discount rate needed to rationalize the observed values. Outside a small region with quite low values of ψ and α , the implied volatility of the discount rate is far too high to make sense, even in the light of evidence from financial economics that discounts applied to business income are quite volatile.

2.4 Graphical discussion

Figure 7 illustrates how the model responds to productivity declines and discount increases for different combinations of the parameters ψ and α . All of the graphs show an upward-sloping job creation curve that relates the employer’s margin, $m = x - w$, to market tightness θ . It is

$$m = (r + s)xcT(\theta). \tag{16}$$

An increase in the discount rate r shifts the job-creation curve upward—to achieve a given level of tightness, employers require a higher margin to engage in the corresponding level of recruiting effort. A decrease in productivity x shifts the curve downward, but for realistic changes in productivity, this effect is almost invisible.

The graphs also show a downward-sloping wage-determination curve,

$$m = (1 - \beta)(x - z(x; \alpha) - \beta c s \hat{\theta}(\theta; \psi)). \tag{17}$$

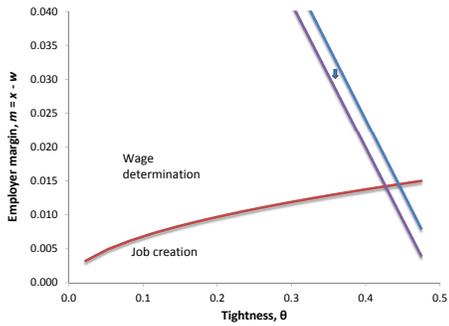
Here $z(x; \alpha) = [(1 - \alpha)x + \alpha\bar{x}]\bar{z}$ and $\hat{\theta}(\theta; \psi) = \psi\theta + (1 - \psi)\bar{\theta}$. A decline in productivity shifts this curve downward. A rise in the discount does not shift this curve.

In the graphs, the decline in productivity is one percent and the increase in the discount rate is from 10 percent per year to 20 percent per year. Graph (a) describes the model with Nash bargaining and fixed flow value of unemployment, z . A decline in productivity of causes only a small decline in tightness, as in Shimer (2005). The wage curve is so steep that its downward shift has little effect on its horizontal position, reflecting the strength of the negative feedback through the tightness effect on the wage. Graph (b) shows the effect of an increase in the discount rate in the same economy. The upward shift in the job-creation curve has little effect because of the steep wage-determination curve.

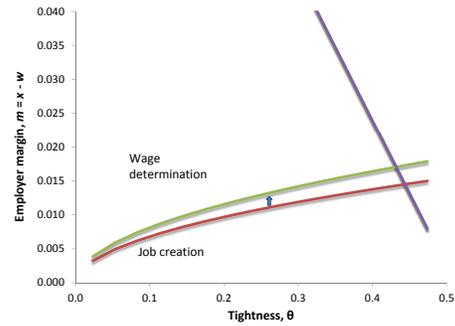
Graphs (c) and (d) describe an economy where wage determination is isolated from tightness ($\psi = 0$). With a flat wage-determination curve, the effects of both a productivity decline and a discount increase are large, as in Shimer (2004).

Graphs (e) and (f) maintain the isolation of wage determination from tightness and add isolation from productivity shifts by setting $\alpha = 0$. In graph (e), neither curve shifts in response to a productivity decline, so nothing happens to tightness. Graph (f) shows a large decline in tightness—the same as in graph (d)—from a rise in the discount rate. This specification achieves what seems to be necessary to make sense of the low correlation of productivity and tightness yet maintain a coherent explanation of the high volatility of tightness.

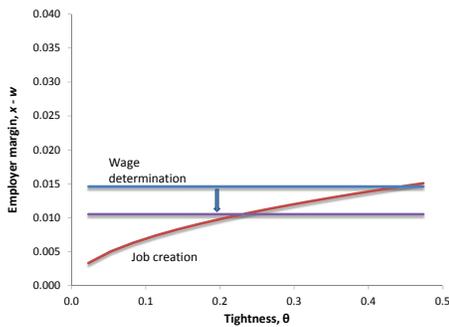
Figure 8 lays out the parameter space graphically. The upper right-hand quadrant includes the DMP model as formulated in Mortensen and Pissarides (1994). With ψ close to one, Nash bargaining determines wages. With α close to one, the flow value of unemployment, z , is fixed, so the gap between productivity and that flow value is sensitive to productivity, and thus tightness depends on productivity. If these parameter values are correct, the implied volatility of the discount rate should correspond to beliefs and evidence about that volatility. If the parameters are wrong—for example, if Shimer (2005) is correct that the movements of tightness are far too large for consistency with that version of the DMP model—implausibly large movements of the implied discount rate will occur for parameter values in the upper-right quadrant. Shimer showed that tightness was hardly sensitive at all to movements in productivity in the Mortensen and Pissarides (1994) model. As I noted earlier, his point carries over to movements in the discount rate. If it takes huge movements in the discount rate to explain the observed volatility of tightness, a calculation of the implied discount rate given parameter values in the upper-right quadrant will have



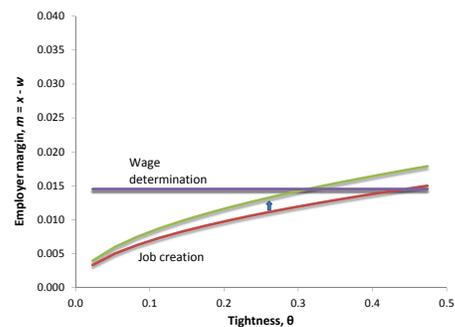
(a) Nash: $\psi = 1$ and $\alpha = 1$, productivity decline



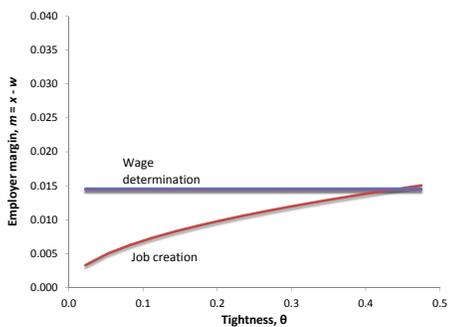
(b) Nash: $\psi = 1$ and $\alpha = 1$, discount increase



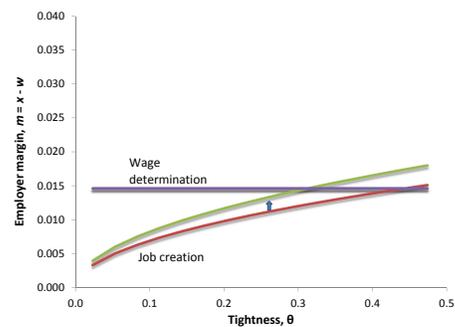
(c) Tightness insulation: $\psi = 0$ and $\alpha = 1$, productivity decline



(d) Tightness insulation: $\psi = 0$ and $\alpha = 1$, discount increase



(e) Tightness and productivity insulation: $\psi = 0$ and $\alpha = 0$, productivity decline



(f) Tightness and productivity insulation: $\psi = 0$ and $\alpha = 0$, discount increase

Figure 7: Effects of Shifts in Productivity and Discount Rate for Combinations of Parameter Values

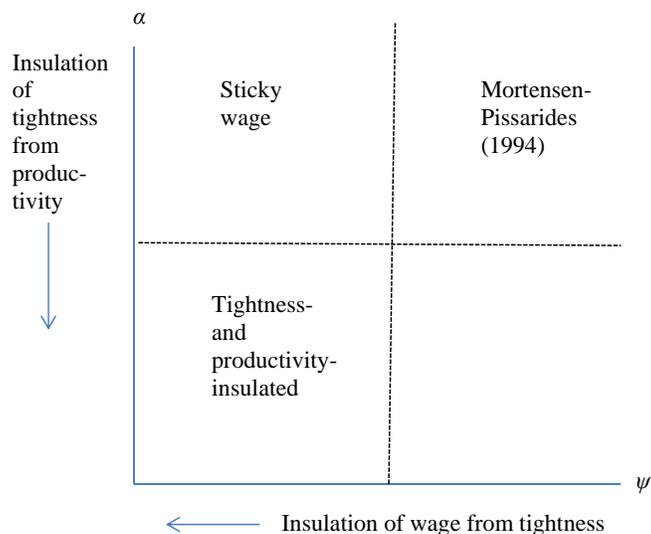


Figure 8: DMP Models within the Parameter Space

huge volatility. The finding of high volatility in that quadrant is a restatement of Shimer’s point.

The upper-left quadrant of Figure 8 describes sticky-wage models. In the decade since Shimer’s finding altered the course of research in the DMP class of models, many rationalizations of sticky wages have appeared—way too numerous to list here. Many achieved the needed stickiness by limiting the response of the wage to labor-market tightness, as in this quadrant. Again, if the hypothesis of a low value of ψ and the standard view that z does not change over the cycle as productivity changes are correct, implied discount rates made with these parameter values should be reasonably but not excessively volatile. If, on the other hand, the wage stickiness associated with a low value of ψ exaggerates the effect of productivity on tightness, then the implied discount rate will be unreasonably volatile because it will move to offset the exaggerated effects of productivity. Thus a finding of a high volatility of the implied discount rate will point away from a standard sticky-wage model.

Finally, the lower-left quadrant of Figure 8 contains DMP models that respond weakly to productivity movements but are reasonably sensitive to movements in discount rates. These models are sensitive to the gap $x - z$ that is the basic source of the response of tightness to productivity movements, but the gap hardly changes when x changes because z changes in parallel. If these parameter values hold, the implied time series for the discount rate will be correct, so its volatility will be in line with evidence and beliefs. On the other hand, if, for

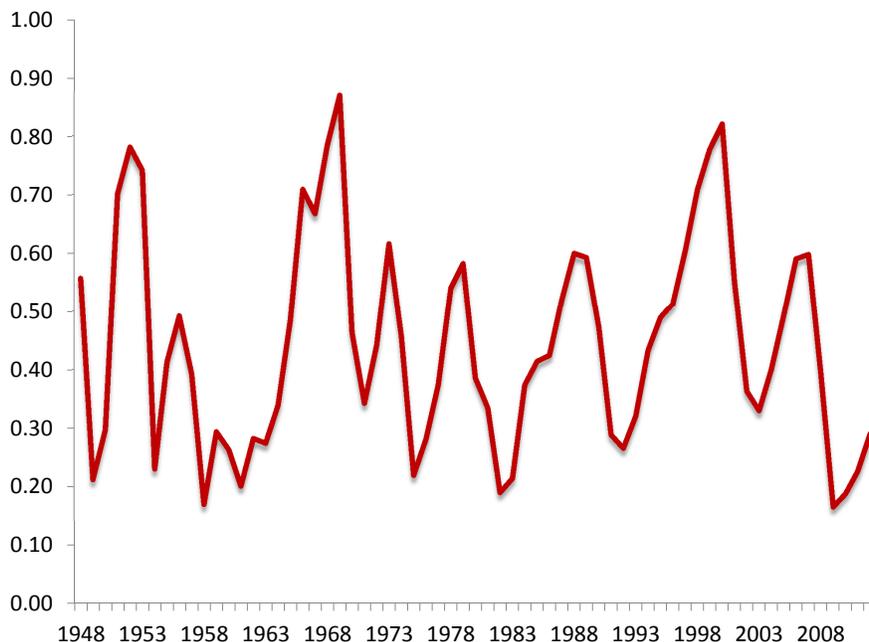


Figure 9: The Vacancy/Unemployment Ratio, θ , 1948 through 2012

example, a low value of α understates the role of productivity movements on tightness, the implied discount rate will make up for the neglected influence of productivity and will have an implausible volatility.

3 Measuring the Implied Volatility of the Discount Rate

3.1 Data

I use annual data for 1948 through 2012. JOLTS measures the stock of vacancies. I divide the number of vacancies in all sectors including government (BLS series JTU00000000JOL) by the number of unemployed workers (BLS series LNS13000000), to obtain θ for the years after 2000. For the earlier years, Petrosky-Nadeau and Zhang (2013) have compiled data on the job vacancy rate beginning in 1929. For the years before 2001, I take the ratio of their vacancy rate to the unemployment rate as a proxy for θ , which I rescale to match the JOLTS-based estimates of θ during the later years. The resulting series for θ has a downward trend, reflecting declining matching efficiency. I remove the trend with a regression of $\log \theta_t$ on a time trend and restate earlier years at the average level of recent years. Figure 9 shows the resulting series—it is a reliable and consistent cyclical indicator.

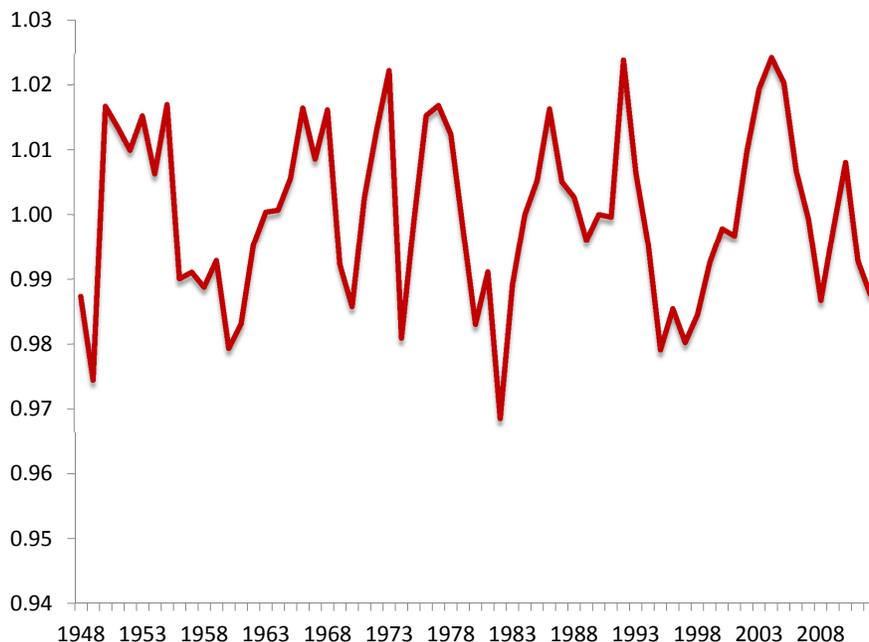


Figure 10: Labor Productivity, 1948 through 2012

With a Cobb-Douglas technology, the marginal product of labor is the output/labor ratio multiplied by the elasticity of the production function with respect to labor. Productivity growth has substantial medium- and low-frequency components that arguably do not have the same effects as cyclical movements, possibly because non-market productivity moves in the same way as market productivity at lower frequencies. I follow Shimer (2005) in removing a Hodrick-Prescott trend from the data on the output/labor ratio (BLS series PRS84006093). Figure 10 shows the movements in the resulting series. Note that the deviations from normal are quite small—the log-standard deviation of the series is only 1.3 percent in annual data.

A key fact about θ and x is their low correlation, 0.18. Plainly x is not the sole determinant of labor-market tightness. Figure 11 shows the scatter plot of the two variables from 1948 through 2012. The low correlation is not the result of a highly nonlinear close relationship, but must be the result of other influences, notably shifts in the discount rate.

3.2 Results on implied volatility of the discount rate

For any point in the (α, ψ) space, the time series r_t that solves the equilibrium condition,

$$\frac{(1 - \beta)(x_t - z_t) - \beta x_t c \hat{\theta}_t}{r_t + s} = c x_t \frac{\sqrt{\theta_t}}{\mu} \quad (18)$$

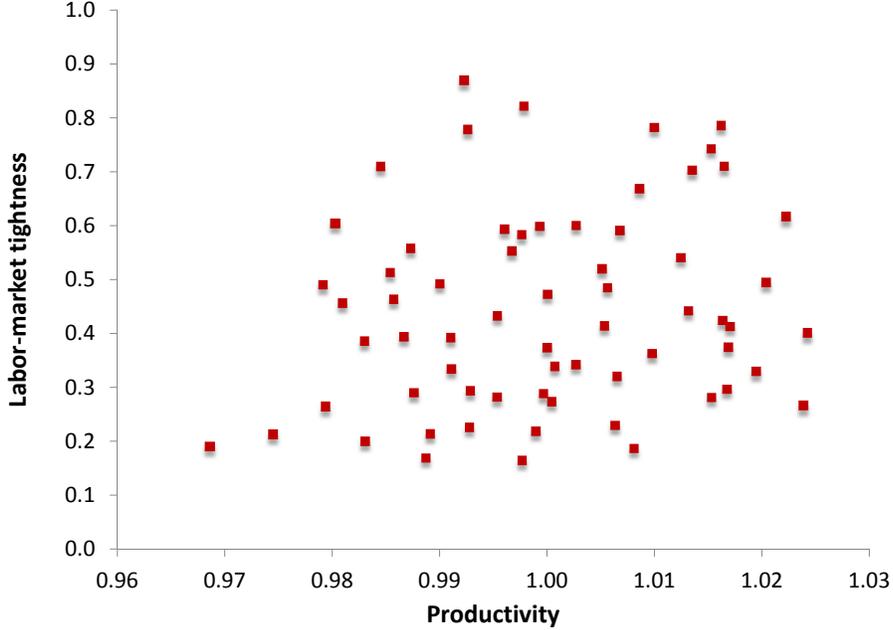


Figure 11: Tightness θ and Productivity x , 1948 through 2012

with $z_t = [\alpha\bar{x} + (1 - \alpha)x_t]\bar{z}$ and $\hat{\theta}_t = \psi\theta_t + (1 - \psi)\bar{\theta}$, is the discount rate that accounts for the values of tightness θ_t and productivity x_t in each year. For example, if the labor market is tight, with a low θ_t in a year when productivity is not unusually high, the calculation infers that a low discount rate accounts for employers' enthusiasm in recruiting workers.

Calculating the expected return or discount rate for investment in new workers is analogous to calculating the expected return to investment in physical capital, such as in Hall (2013a).

Table 1 shows the implied standard deviations of the discount rate at many points in the parameter space, stated as percents at annual rates. Those shaded green correspond to evidence discussed later in this paper about the likely volatility of the discount rate and those more toward the red shade are too high to be reasonable in the light of that evidence. The results reject values of ψ above 0.1. The evidence in favor of stickiness in the sense of isolation of wages from tightness is strong. The evidence favors the markdown hypotheses—that the market/non-market flow gap $x - z$ hardly responds to productivity—the standard deviation of 33 percent for $\alpha = 1$ and $\psi = 0$ is much higher than the standard deviation of 20 percent for $\alpha = 0$ and $\psi = 0$. The calculations suggest that there is no difference among values of α below around 0.5.

		ψ : weight on tightness in wage determination										
		0	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1
α : size of constant in non-market value	1	22	30	43	57	71	86	101	115	130	146	161
	0.8	18	28	42	56	71	86	101	116	131	146	161
	0.6	15	27	41	56	71	86	101	116	132	147	162
	0.4	13	27	42	57	72	87	102	117	132	148	163
	0.2	12	27	42	57	72	88	103	118	133	149	164
	0	12	28	43	58	73	89	104	119	134	150	165

Table 1: Standard Deviations of Implied Discount Rates within the Parameter Space, Percents at Annual Rates

What the calculations show is that productivity is not a good candidate as a driving force for tightness. This conclusion flows from the lack of anything like a systematic relation between productivity and tightness, as shown in Figure 11. The calculation of the implied discount creates a driving force that, by construction, does a good job of explaining the movements of tightness. The ultimate test is whether the implied discount rate resembles discount rates constructed from other sources.

The appendix presents a version of the credible-bargaining model of Hall and Milgrom (2008). Its version of Table 1 has quite similar results supporting the conclusion in favor of low values of α and of the parameter that is equivalent to ψ .

Early in the paper, I noted that recruiting cost may involve a fixed component in addition to the component that is proportional to the duration of a vacancy. In the presence of a fixed component, the implied volatility of the discount is lower. For example, if the fixed recruiting cost is 0.10, about a third of the total cost per recruit of 0.33 ($J = 0.46$, $\kappa = 0.10$, and $\tilde{J} = 0.56$), the standard deviation of the annual discount with $\alpha = 0$ and $\psi = 0$ is 8 percent annually, in place of the 12 percent for $\kappa = 0$.

4 Discount Rates in the Stock Market

Finance theory defines the discount ratio, $\frac{1}{1+r}$, as the ratio of the current market price of a future cash receipt to the statistical expected value of the future receipt. The quantity r is the discount rate. An intuitive but not quite obvious result of finance theory is that the discount rate for a particular future cash flow is the expected rate of return to holding a claim to the cash flow. Note that discount rates are specific to a future cash flow—the

discount rate for a safe cash flow, one paying as much in good times as in bad times, is lower than for a risky cash flow, one paying more in good times than in bad times. The discount rate may be negative for a cash flow with insurance value, one paying less in good times than in bad times. The discount rate reflects the risk premium associated with a future cash flow.

This paper does not explain why risky flows receive higher discounts in recessions (but see Bianchi, Ilut and Schneider (2012) for a new stab at an explanation). Rather, it documents that fact by extracting the discount rates implicit in the stock market.

4.1 The discount rate for the S&P stock-price index

The issue of the expected return or discount rate on broad stock-market indexes has received much attention in financial economics since Campbell and Shiller (1988). Cochrane (2011) provides a recent discussion of the issue. Research on this topic has found that two variables, the level of the stock market and the level of consumption, are reliable forecasters of the return to an index such as the S&P. In both cases, the variables need to be normalized. Figure 12 shows the one-year ahead forecast from a regression where the left-hand variable is the one-year real return on the S&P and the right-hand variables are a constant, the log of the ratio of the S&P at the beginning of the period to its dividends averaged over the prior year, and the log of the ratio of real consumption to disposable income in the month prior to the beginning of the period. The graph is quite similar to Figure 3 in Cochrane's paper for his equation that includes consumption.

The standard deviation of the discount rate in Figure 12 is 7.2 percentage points at an annual rate. This is an understatement of the true variation, because it is based on an econometric forecast using only a subset of the information available at the time the forecasts would have been made.

Another source of evidence on the volatility of expected returns in the stock market comes from the Livingston survey, which has been recording professional forecasts of the S&P stock-price index since 1952. The standard deviation of the one-year forward expected change in the index in real terms plus the current dividend yield is 5.8 percent.

So far I have considered the volatility of the expected return in the stock market for an investment held for one year. The future cash flow subject to discount is the value from selling the stock in a year, inclusive of the dividends earned over the year reinvested in the

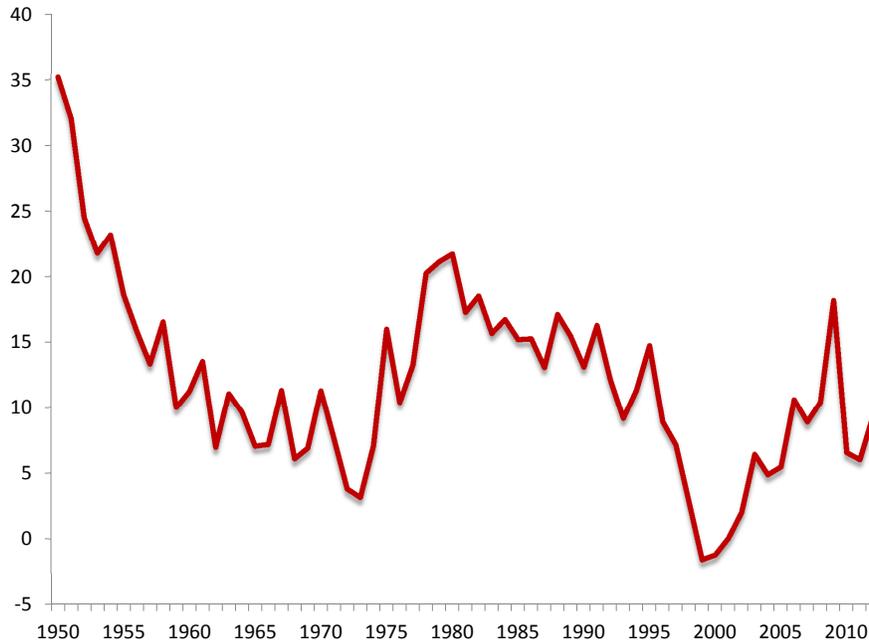


Figure 12: Econometric Measure of the Discount Rate for the S&P Stock-Price Index

same stock. Most of the risk arises from fluctuations in the price of the stock rather than from the value of the dividends, so the risk under consideration in calculating the expected return arises from all future time periods, not just from the year of the calculation. The stock market looks much further into the future than does a firm evaluating the benefit from hiring a worker, as most jobs last only a few years. One way to deal with that issue is to study the valuation of claims to dividends accruing over near-term intervals. Such claims are called “dividend strips” and trade in active markets. Because dividends are close to smoothed earnings, values of dividend strips reveal valuations of near-term earnings. Jules van Binsbergen, Brandt and Kojen (2012) and van Binsbergen, Hueskes, Kojen and Vrugt (2013) pioneered the study of the valuation of dividend strips, with the important conclusion that the volatility of discount rates for near-term dividends is comparable to the volatility of the discount rate for the entire return from the stock market over similar durations.

These authors study two bodies of data on dividend strips. The first infers the prices from traded options. Buying a put and selling a call with the same strike price and maturity gives the holder the strike price less the stock price with certainty at maturity. Holding the stock as well means that the only consequence of the overall position is to receive the intervening dividends and pay the riskless interest rate on the amount of the strike price. The second source of data comes from the dividend futures market. The latter provides data

for about the last decade, whereas data from options markets are available starting in 1996. Jules van Binsbergen et al. (2012) published the options-based dividend strip data on the AER website, for six-month periods up to two years in the future.

The market discount rate for dividends payable in 13 through 24 months is

$$r_t = \frac{\mathbb{E}_t \sum_{\tau=13}^{24} d_{t+\tau}}{p_t} - 1, \quad (19)$$

where d_t is the dividend paid in month t and p_t is the market price in month t of the claim to future dividends inferred from options prices and the stock price. Measuring the conditional expectation of future dividends in the numerator is in principle challenging, but seems not to matter much in this case. I have experimented with discount rates for two polar extremes. First is a naive forecast, taking the expected value to be the same as the sum of the 6 most recently observed monthly dividends as of month t . The second is a perfect-foresight forecast, the realized value of dividends 13 through 24 months in the future. The discount rates are very similar. Here I use the average of the two series.

The main point of van Binsbergen et al. (2012) and van Binsbergen et al. (2013) is that the discounts (expected returns) embodied in the prices of near-future dividend strips are remarkably volatile. Many of the explanations of the volatility of expected returns in the stock market itself emphasize longer-run influences and imply low volatility of near-term discounts, but the fact is that near-term discount volatility is about as high as overall discount volatility. In the earlier years, some of the volatility seems to arise from pricing errors or noise in the data. For example, in February 2001, the strip sold for \$9.37 at a time when the current dividend was \$16.07 and the strip ultimately paid \$15.87. The spike in late 2001 occurred at the time of 9/11 and may be genuine. No similarly suspicious spikes appear in the later years.

Over the period when these authors have compiled the needed options price, from 1996 through 2009, the standard deviation of the market discount rate on S&P500 dividends to be received 13 to 24 months in the future, stated at an annual rate in real terms, is 10.1 percent. The standard deviations of the discount rate for the stock market over the same period are 5.4 percent for the econometric version of the return forecast and 6.2 percent for the return based on the Livingston survey.

Figure 13 shows the three series for the discount rates implicit in the S&P stock price and in the prices of dividend strips for that portfolio. On some points, the three series agree, notably on the spike in the discount rate in 2009 after the financial crisis. In 2001, the

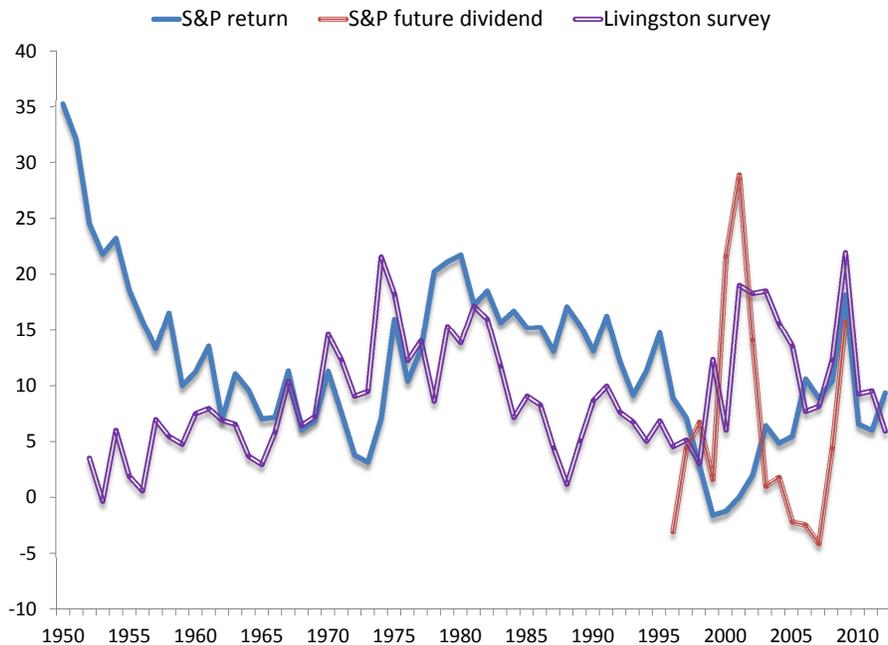


Figure 13: Three Measures of Discount Rates Related to the S&P Stock Price Index Portfolio

<i>Measures</i>	<i>Correlation</i>	<i>Years</i>
Dividends, stock price	-0.32	1996-2009
Dividends, Livingston	0.37	1996-2009
Stock price, Livingston	-0.14	1952-2012

Table 2: Correlations among the Three Measures of Discount Rates

Livingston forecasters and the strips market revealed a comparable spike, but the econometric forecast disagreed completely—high values of the stock market and consumption suggested a low expected return. From 1950 to 1960, the reverse occurred. The Livingston panel had low expectations of a rising price, whereas the econometric forecast responded to the low level of the stock price relative to dividends, normally a signal of high expected returns. Table 2 shows the correlations of the three measures.

The three measures of discount rate related to the S&P portfolio all have similar volatility, in the range from 6 to 10 percent at annual rates. Contrary to expectation, the three are not positively correlated. Two of the three correlations are negative, though measured over a brief and partly turbulent period. Finance theory imposes no restrictions on the correlations of discount rates for different claims on future cash, because the discounts incorporate risk premiums that may change over time in different ways for different claims. Explaining

<i>Measure</i>	<i>Correlation with labor market</i>	<i>Years</i>
Dividends	0.14	1996-2009
Stock price	0.18	1950-2009
Livingston	0.29	1952-2012

Table 3: Correlations of the Discount Rate in the Labor Market with Stock-Market Rates

the dramatic differences between regression-based measures of expected returns and those obtained from surveys of experts about the same expected returns involves many other considerations about the limitations of the information available to the econometrician, biases from specification search, and the use of information not available to market participants, together with questions about the reliability of an expert panel’s forecasts if they are not actively involved in trading the portfolio.

For this paper, the key conclusions from this review of financial discount rates are, first, their fairly high volatilities, and, second, their low correlations with each other. In view of those low and negative correlations, it would not be realistic to adopt any one of the measures derived from the stock market and plug it into the DMP model.

5 The Plausibility of the Calculated Discount Rate for Hiring

Table 3 shows the correlations of the labor-market discount rate corresponding to the preferred case of $\alpha = 0$ and $\psi = 0$, as discussed in Section 3, and the three discounts inferred from the stock market. By the standard of the correlations among those three, the correlations with the labor market are impressive. Figure 14 plots the time series for the labor-market r_t and the discount rate derived from the Livingston forecasts. The relation between the two is quite apparent after 1970. Spikes in both discount rates occur regularly in recessions, in 1970, 1974-75, 1980-82 (two recessions), 1990, 2001, and 2008-09.

The expected return in the labor market rises much more in response to an adverse shock than does the expected return in the stock market. As Figure 4 shows, the job value and stock price have roughly equal volatility. The job value is the expected flow value of a new worker discounted by the expected return in the labor market; the stock price is the

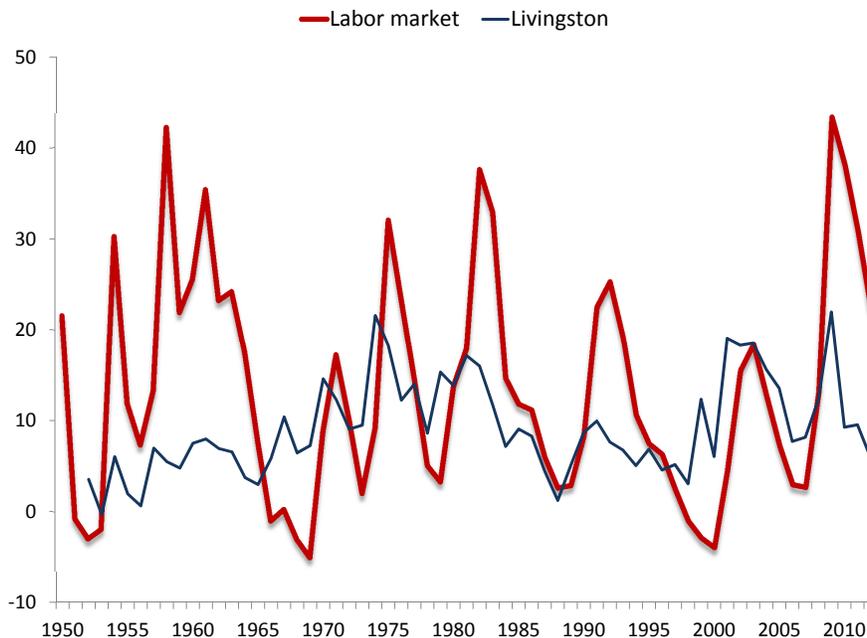


Figure 14: Discount Rate for the Labor Market and the Livingston Panel's Rate for the Stock Market

expected flow of dividends or profit discounted by the expected return in the stock market. The finding that α is fairly low implies that the flow value of a new worker is quite stable, because, as Figure 10 shows, the marginal product of labor is cyclically stable—much more stable than corporate profits or dividends. Hence the volatility of the discount rate in the labor market must be greater than the volatility of the discount rate for profits or dividends to result in the roughly equal volatilities of the two discounted values. The higher volatility of the labor-market discount implies a larger increase in the financial risk of the flow value of a worker in recessions than the rise in the financial risk of corporate income in general.

I conclude that financial economics confirms the visual impression in Figure 4 that the same principles influence the valuation of the employer's net flow value from an employment relationship as influence the stock market's valuation of corporate profits. Thus events such as a financial crisis increase the discount rates applied to both flows. In the labor market, employers respond by cutting back on job creation because the capital value of a new employment relationship is the driving force for job creation. The data support this view without any stretch in terms of parameter values.

The DMP framework provides a robust way to measure the resulting discounted value, the job value J , because it is the cost of recruiting a new worker. A direct comparison of

J with the value of the U.S. stock market shows a remarkable similarity over the past two decades. The comparison based on finance theory confirms what the naked eye sees—that the job value fell after the financial crisis in line with the discounts implicit in the stock market. To put it differently, a large increase in the equity premium appears to have applied to the net benefit of hiring a new worker as well as to the stock market.

6 Evidence on the Relationship between Productivity and the Flow Value of Unemployment

Chodorow-Reich and Karabarbounis (2013) study the movements of the flow value of unemployment, z , or, as they term the concept, the opportunity cost of work. I believe that theirs is the only contribution on this subject. They conclude that z moves at cyclical frequencies in proportion to productivity. They start from an expression for z in Hall and Milgrom (2008) with two components. The first is the level of unemployment-conditioned benefits. The second measures the remaining impact on family welfare from shifting a worker from non-market to market activities, counting the lost flow value of home production, the increase in consumption allocated to workers compared to non-workers, and the change in flow utility for a family member moving from non-market activities to market work. Though unemployment insurance rises dramatically in recessions, some of which occur at the same time as drops in productivity, a decline in other types of unemployment-conditioned benefits offsets this effect. And the second component rises in proportion to productivity. Their research supports the hypothesis that the wage is a markdown on productivity because $x - z$ is stable at cyclical frequencies.

7 Direct Measurement of the Flow Value of a New Worker

This paper uses the model of Section 2 to infer w from x , and thus to infer the flow value of a new worker, $x - w$. A natural question is the feasibility of a direct attack on the measurement of the flow value. Why not measure w from data on employee earnings instead? I am skeptical that such an approach would work. Most of workers' earnings are Ricardian rents to a primary factor. Thus the gross benefit of a new hire is just a bit higher than the

wage. The difference between a necessarily noisy measure of the gross benefit (the product) and the wage, also measured with noise, would be almost entirely noise.

Suppose that a newly hired worker faces a constant monthly hazard $s = 0.035$, of separation, as documented earlier, and the discount rate applicable to the financial risk of $x - w$ is 10 percent per year or 0.0083 per month. The capitalization factor for a monthly flow is then

$$\frac{1}{0.035 + 0.0083} = 23 \tag{20}$$

From Figure 1, the decline in the job value that occurred in the Great Recession was about \$300. Thus the decline in the monthly net flow to the employer, $x - w$, was $300/23 = \$13$ per month. The median hourly wage in 2011 was \$17, so the decline in the monthly flow was equal to about 45 minutes of wage earnings or a fraction 0.005 of monthly full-time earnings.

Because the change in the net flow value of a newly hired worker needed to rationalize the observed increase in unemployment following the crisis is absolutely tiny compared to earnings and other flows, it appears hopeless to measure the job value by determining the flow and calculating its capital value to the employer. Haefke, Sonntag and van Rens (2012), in an ambitious attempt to estimate the response of wages to productivity, concluded that it was not possible to pin it down with a sufficiently small standard error to resolve the subtle question of the cyclical variability of the flow.

Other reasons that direct measurement of the flow value of worker is impractical are (1) training costs are a deduction from productivity and so need to be measured separately, and (2) as Yashiv (2013) observes, labor adjustment costs are a deduction from the flow value.

8 Concluding Remarks

The suggestion in this paper that the discount rate is a driving force of unemployment is not new. In addition to the work of Pissarides, Yashiv, and Merz in the DMP framework already mentioned, Phelps (1994), pages 61 and 171, considers the issue in a different framework. Mukoyama (2009) is a more recent contribution focusing on discount volatility in the stock market. Still, most recent research in the now-dominant DMP framework concentrates on productivity as the driving force. The conclusion of this paper with respect to fluctuations in productivity is rather different. Because the evidence favors sticky wages in the sense of almost complete insulation of the wage from tightness, if productivity fell by one or two percentage points while the flow value of unemployment, z , remained unchanged, unemploy-

ment would rise sharply. But the conclusion of the paper and the work of Chodorow-Reich and Karabarbounis (2013) is that z falls in proportion to productivity, implying that such a decline in productivity has little effect on unemployment. Sticky wages co-exist with small responses of unemployment to the modest changes in productivity that occur in the U.S. economy. I believe that no researcher has tried to make the case that any actual decline in productivity occurred following the financial crisis is anywhere near large enough or timed in the right way to explain the high and lingering unemployment rate in the U.S., much less in countries like Spain where unemployment rose into the 20-percent range.

The novelty in this paper is its connection with the finance literature that quantifies the large movements in the discount rates in the stock market. This literature has reached the inescapable conclusion that the large movements in the value of the stock market arise mainly from changes in discount rates and only secondarily from changes in the profit flow capitalized in the stock market. The field is far from agreement on the reasons for the volatility of discount rates.

In view of these facts, it is close to irresistible to conclude that whatever forces account for wide variations in the discount rates in the stock market also apply to the similar valuation problem that employers face when considering recruiting. If so, even the highly stable net flow value of a worker found in this paper generates fluctuations consistent with the observed large swings in unemployment.

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Appendix

A Related Research

Research in the DMP framework has considered three driving forces, apart from productivity. The first is that declines in product demand cause firms to move down their marginal cost curves. The firms have sticky prices, so the marginal revenue product of labor falls. The consequences in the DMP model are then the same as for a decline in productivity. The second channel involves declining price inflation. If the bargain between a newly hired worker and an employer involves an expected rise in the nominal wage that is sticky, but the growth of prices falls to a lower level, the benefit of a new hire to an employer falls and unemployment rises, according to standard DMP principles. The third channel invokes increases in the flow value of unemployment, z , on account of more generous unemployment insurance benefits.

A.1 Sticky prices

Walsh (2003) first brought a nominal influence into the DMP model. Employers in his New Keynesian model have market power, so the variable that measures the total payoff to employment is the marginal revenue product of labor in place of the marginal product of labor in the original DMP model. Price stickiness results in variations in market power because sellers cannot raise their prices when an expansive force raises their costs, so the price-cost margin shrinks. Rotemberg and Woodford (1999) give a definitive discussion of the mechanism, but see Nekarda and Ramey (2010) for negative empirical evidence on the cyclical behavior of margins. Hall (2009) discusses this issue further. The version of the New Keynesian model emphasizing price stickiness suffers from its weak theoretical foundations and has also come into question because empirical research on individual prices reveal more complicated patterns with more frequent price changes than the model implies. Hall (2013b) finds evidence against higher margins in slumps because advertising should rise substantially with margins but in fact advertising falls dramatically in recessions.

Walsh adopts the Nash wage bargain of the canonical DMP model, which implies that his model may generate low unemployment responses for the reason that Shimer (2005) pointed out. Conceptually, it remains the case that Walsh was the first to resolve the clash between Keynesian models with excess product supply and the DMP model of unemployment.

Mortensen (2011) establishes a direct connection between drops in product demand and the payoff to new hires. He makes the simple assumption that firms stick to their earlier prices when demand drops, so that firms are quantity-takers. He uses a Dixit-Stiglitz setup to map the consequences back into the labor market and shows that the fixity of output results in potentially large declines in the net benefit of a new hire.

A.2 Sticky wages

Another approach introduces a nominal element into wage determination. The canon of the modern New Keynesian model, Christiano, Eichenbaum and Evans (2005), has workers setting wages that are fixed in nominal terms until a Poisson event occurs, mirroring price setting in older versions of the New Keynesian model. That paper does not have a DMP labor market. Gertler, Sala and Trigari (2008) (GST) embed a DMP labor-market model in a general-equilibrium model, overcoming Shimer’s finding by replacing Nash bargaining at the time of hire with a form of wage stickiness. Gertler and Trigari (2009) developed the labor-market specification. A Poisson event controls firm-level wage bargaining, which takes the Nash form. Between bargaining times, the wage of newly hired workers adheres to the most recent bargain. If labor demand turns out to be higher than expected at bargaining time, the part of the surplus captured by the employer rises and the incentive to recruit workers rises. By standard DMP principles, the labor market tightens and unemployment falls. Though the model is Keynesian in the sense of sticky wages, it describes an equilibrium in the labor market in the sense of Hall (2005)—the relation between workers and an employer is privately efficient. GST build a model of the general-equilibrium response to monetary and other shocks in a version of the Gertler-Trigari setup where the wage bargain is made in nominal terms. The GST paper resolves the clash by making the DMP determination of unemployment sensitive to the rate of inflation.

A key idea in Gertler and Trigari (2009), put to work in the GST paper, is that workers hired between bargaining times inherit their wage terms from the most recent bargain. In principle, this setup could violate the private efficiency criterion by setting the wage too high to deliver a positive job value to the employer or too low to deliver a job value below the job candidate’s reservation level, but, again, in practice this is not likely to occur. If it were an issue, the introduction of state-dependent bargaining would solve the problem, at the cost of a more complicated model.

The GST model assumes that the wage bargain is made in money terms, as the traditional Keynesian literature likes to say. The substance of the assumption is that a state variable—the most recently bargained nominal wage—influences the job value for new hires until the next bargain occurs. This assumption has had a behavioral tinge in that literature—the role of the stale nominal wage arises from stubbornness of workers or employers or from money illusion. From the perspective of bargaining theory, however, as long as the stale wage keeps the job value in the bargaining set, that wage is an eligible bargain. See Hall (2005) for further discussion, not specifically in the context of a nominal state variable. There’s no departure from strict rationality in the GST model.

The implications of a model linking the current job value to a stale nominal variable are immediate: The more the price level rises from bargaining time to the present, the higher is the job value in real terms.

A.2.1 Unemployment insurance

Recent papers by Nakajima (2012), Valletta and Kuang (2010), Fujita (2011), and Daly, Hobijn, Şahin and Valletta (2011), culminating in Farber and Valletta (2013), ask whether higher UI benefits result in lower search effort and higher reservation wages, both of which would raise unemployment in a standard DMP model. This research compares the job-finding rates of covered workers to uncovered workers. The answer is fairly uniformly that the effects of UI enhancements during times of high unemployment in raising unemployment still further are quite small, in the range of 0.3 percentage points of extra unemployment.

Hagedorn, Karahan, Manovskii and Mitman (2013) tackle a more challenging question, whether more generous UI benefits result in higher wages and higher unemployment by raising the flow value of unemployment z and thus shrinking the $x - z$ gap that measures the fundamental force that determines the unemployment rate. They compare labor markets with arguably similar conditions apart from the UI benefits regime. In their work, the markets are defined as counties and the similarity arises because they focus on pairs of adjacent counties. The difference in the UI regimes arises because the two counties are in different states and UI benefits are set at the state level and often differ across state boundaries. The research uses a regression-discontinuity design, where the discontinuity is the state boundary and the window is the area of the two adjacent counties. The authors conclude that, absent the increase in UI benefits, unemployment in 2010 would have been about 3 percentage points lower.

Many commentators have dismissed Hagedorn and co-authors' conclusion on the grounds that the research implies that unemployment would have hardly risen at all absent the extension of UI benefits, despite the financial crisis and resulting collapse of product demand. But the dismissal is unwarranted. Hagedorn and co-authors' work fully recognizes that the enhancements of UI benefits was itself the result of the forces that caused the Great Recession. The proper interpretation, with the framework of the paper, is that feedback from enhanced UI benefits was a powerful amplification mechanism of a negative impulse arising from the crisis.

The issues that arise in evaluating the paper are those for any regression-discontinuity research design: (1) Are there any other sources of discontinuous changes at the designated discontinuity points that might be correlated with the one of interest? (2) Is the window small enough to avoid contamination from differences that do not occur at the discontinuity point but rather elsewhere in the window? The authors explore a number of state-level economic policies that could generate cross-border effects that might be correlated with the UI effects, but none seem to matter. The authors are less persuasive on the second point. Many counties are large enough to create substantial contamination. Far from being atoms, single counties are often large parts of their states, both geographically and in terms of the share of the population. The extreme case is Washington, DC, treated as a state with only one county.

Hagedorn and co-authors have ignited an important debate. Further discussion may help resolve the issue of the reliability of their finding of such large effects on wages and unemployment.

A.3 Forming the present value of a newly hired worker's net benefit to the employer

Yashiv (2000) undertook the task I declare to be impossible in the body of this paper: forming the present value of the difference between a worker's marginal product and wage. Equation (4) in his paper is equivalent to equation (6) here. On page 492, Yashiv notes the analogy between the valuation of a worker's net contribution and valuation in the stock market of a stream of dividends. One important difference between his approach and mine is that he takes the hiring cost to be strongly convex in the flow of hires at the level of the firm whereas I adopt the linearity that is that standard property of the DMP class of models.

Under linearity, the asset value of the employment relationship is observed directly. By contrast, Yashiv uses GMM to infer the marginal hiring cost. A second important difference is that Yashiv’s approach does not distinguish between hiring costs incurred prior to a wage bargain and those following the bargain.

A.4 Variations in discounts

The basic proposition that the stock market varies largely because of changes in discount rates is the conclusion of a famous paper, Campbell and Shiller (1988). Cochrane (2011) discusses the finding extensively.

Greenwald, Lettau and Ludvigson (2014) decompose movements of expected stock-market returns and assign a large fraction of the variation to shifts in discounts not affiliated with other variables, such as consumption growth.

Gourio (2012) builds a model where a small probability of a disaster generates substantial variations in discounts, which influence employment through a standard labor-supply setup. The model does not include unemployment.

A.5 Joint movements of labor-market variables and the stock market

Kuehn et al. (2013) is an ambitious general-equilibrium model that combines a DMP labor market with a full treatment of financial markets. Its goal is roughly the reverse of the goal of this paper. It makes the case that volatility in real allocations resulting from amplification of productivity shocks in the labor market causes financial volatility. In particular, the model can generate episodes that look like financial crises, with dramatic widening of the equity premium. The paper provides an endogenous source of economic disasters, an advance over the existing literature that takes large declines in output and consumption to be the result of exogenous collapses of productivity.

Kuehn and coauthors build in a number of the ideas from the post-Shimer literature to gain high amplification in the labor market from productivity shocks. These include (1) adding a fixed cost to the pre-bargain recruiting cost, on top of the cost that varies with the time required to fill a job, (2) assigning the worker a tiny bargaining weight, and (3) assigning a high value to the worker’s activities while unemployed, apart from the value of search. They also build in ideas from modern finance that generate a high and variable equity premium

along with a low and stable real interest rate. These are (1) an extremely high coefficient of relative risk aversion and (2) a quite high elasticity of intertemporal substitution. The paper briefly surveys related earlier contributions linking asset-price volatility to unemployment volatility: Danthine and Donaldson (2002), Uhlig (2007), Gourio (2007), and Favilukis and Lin (2012).

Farmer (2012) noted the relationship between the levels of the stock market and unemployment. He adopts the traditional view that unemployment is simply the difference between labor supply and demand, thus sidestepping the issues considered in this paper.

Merz and Yashiv (2007) study investment, hiring, and the stock market jointly. Adjustment costs for both inputs result in values of Tobin's q for both inputs. They estimate a three-equation system comprising dynamic first-order conditions for investment and hiring and the equality of the market value of the firm to its capital stock and employment level valued by their respective qs . They find a high correlation of their fitted value of the U.S. corporate sector with the actual value.

Hall (2001) considered the same evidence about the market value of the corporate sector as Merz and Yashiv (they adopted my data for the value), but reached rather a different conclusion. My paper rejected the assumption that the value arises only from rents associated with investment adjustment costs. It entertained the hypothesis that the corporate sector acquired highly valuable intangible capital during the run-up of the stock market in the 1990s. The value of that inferred intangible capital collapsed between the writing of the first version of the paper and its appearance in print.

The relation between Merz and Yashiv's work and the approach in this paper is that they rely on the strong assumption that the market value of a firm arises solely from its investments in plant, equipment, and employees. This paper makes the weaker assumption that corporate profits arise from many sources, including its capital stocks, and uses evidence about how the stock market discounts the profit stream to rationalize the observed value of one element of the one part of the profit flow, that arising from the pre-bargain investment that employers make in recruiting workers.

Yashiv (2013) extends his earlier work using a similar specification for joint adjustment costs of investment and hiring. In place of employment levels, the specification uses hiring flows, capturing gross rather than net additions to employment. The hiring costs combine a quadratic term and a DMP-style vacancy holding cost. He computes a Campbell-Shiller-

style decomposition of the returns on capital and labor that confirms the importance of variations in discount rates. The paper holds out the promise of helping DMP-type models better characterize the flow value of a newly hired worker to a firm.

B Credible Bargaining with Tightness Insulated from Productivity

The following equations are adapted slightly from Hall and Milgrom (2008):

$$W_J + V = \delta U + (1 - \delta) \left[\bar{z}\hat{x} + \frac{1}{1+r}(W_E + V) \right]. \quad (21)$$

$$\hat{x} = \alpha\bar{x} + (1 - \alpha)x. \quad (22)$$

$$P - W_E = (1 - \delta) \left[-\gamma\hat{x} + \frac{1}{1+r}(P - W_J) \right]. \quad (23)$$

$$2W = W_J + W_E = \frac{1+r}{r+\delta} [\delta U + (1 - \delta)(\bar{z} + \gamma)\hat{x}] + P - V. \quad (24)$$

$$P = x + \frac{1-s}{1+r}P. \quad (25)$$

$$U = \bar{z}\hat{x} + \frac{1}{1+r} [\phi(W + V) + (1 - \phi)U]. \quad (26)$$

$$V = \frac{1}{1+r} [sU + (1 - s)V]. \quad (27)$$

$$\frac{P - W}{1+r}q = cx. \quad (28)$$

Exceptions from the formulation in the Hall-Milgrom paper are: (1) W_E is the employee's wage offer and W_J is the employer's offer, (2) \bar{z} is the normal level of z , when x has its normal level, \bar{x} , and (3) the wage paid, W , is the average of the equilibrium wage offers that would be made if the jobseeker or the employer made the first offer.

Table 4 shows the version of Table 1 for the credible-bargaining model. Except for the left-hand column for $\delta = 0$, the results are quite similar to the earlier ones. At $\delta = 0$, the credible-bargaining model makes tightness extremely sensitive to the driving forces, which

		δ : role of tightness in wage determination										
		0	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1
α : size of constant in non-market value	1	4.E+02	35	54	72	89	103	117	129	141	151	161
	0.8	5.E+02	33	53	72	89	104	117	130	141	152	161
	0.6	5.E+02	31	53	72	89	104	118	130	142	152	162
	0.4	382	30	53	73	90	105	119	131	143	153	163
	0.2	211	30	54	73	91	106	120	132	143	154	164
	0	2	32	55	75	92	107	121	133	145	155	165

Table 4: Implied Volatility of the Discount Rate in the Credible-Bargaining Model

is why the observed volatility of θ can be explained by a discount rate with a standard deviation of only two percent, provided the effects of productivity are shut down completely with $\alpha = 0$. For positive values of α , the volatility of the discount rate is high because the rate needs to offset huge and unrealistic effects of productivity on tightness.