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**Reference-Dependent Effects
of Unemployment on Mental Well-Being**

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Abstract

This paper provides an empirical analysis of reference-dependent effects of unemployment on mental well-being. We show that the negative effect of unemployment on mental well-being depends on expectations about the future employment status.

Several contributions to the literature have shown that the perception of the individual employment status depends on the surrounding unemployment rate. We argue that expectations are a possible link between unemployment rates and the individual employment status regarding changes in mental well-being. Theoretical foundation comes from models for reference-dependent preferences with endogenous reference points. We provide a simple theoretical model to motivate and structure the empirical analysis. Using data from the German Socio-Economic Panel, we estimate a pairwise interacted model for employment status and expectations over two time periods. Life satisfaction is used as a proxy for mental well-being. To identify a causal effect of unemployment, expectations and their interactions on mental well-being, the analysis relies on fixed effects and exogenous entries into unemployment due to plant closures. We confirm the standard result that unemployment has a

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negative effect on mental well-being. Furthermore, the results deliver empirical evidence for reference-dependent effects of unemployment on mental well-being. We find that becoming unemployed unexpectedly is more severe as if the unemployment was expected. Therefore, this paper contributes to the understanding of how mental well-being is affected by unemployment and delivers empirical support for the theoretical models of reference-dependent preference with endogenous reference points determined by expectations.

Keywords: Subjective Well-Being, Unemployment, Reference-Dependence, Reference Points.

JEL classification numbers: C23, D03, D84, I10, I18, J01, J60.

1 Introduction

The relationship of unemployment and health has been amongst others discussed in a series of papers by Ruhm (e.g. 2000, 2003, 2005) who found that unemployment rates are negatively correlated with mortality rates, health care utilization and chronic conditions. Interpreting mortality rates as a proxy for health he concludes that with decreasing macro-economic circumstances health increases. He reasons that people have more time for health increasing activities during recessions but tend to more risky health behavior during economic upswings (smoking, drinking, etc.). When he analyzes the effect of unemployment rates on case-specific mortality rates and specific chronic diseases he finds that only the variation in suicides and mental illness to be procyclical in macro-economic conditions, i.e. suicide rates and the number of mental health problems increase with unemployment rates. He concludes that mental health and mental well-being behave in sharp contrast to physical well-being (Ruhm 2003, p. 655). Therefore, the relationship between mental well-being and economic conditions should be analyzed separately from physical health conditions.

The number of reported mental health problems is steadily increasing in recent years. Health care expenditures caused by mental illnesses are increasing above average compared to expenditures for physical health problems (Statistisches Bundesamt, 2010). As it seems that mental well-being is differently affected by (macro-) economic circumstances than physical health it is of particular importance for health economists to understand what determines mental well-being.

On the individual level Clark and Oswald (1994) established the general result that subjective well-being is negatively affected from unemployment. Winkelmann and Winkelmann (1998) disentangled the negative effect of unemployment on life satisfaction into a pecuniary and a non-pecuniary effect. The non-pecuniary effect is the psychological burden of unemployment that arises in addition to the loss of income that characterizes the economic burden of unemployment. They found the non-pecuniary effect to be much larger than the effect that stems from the associated loss of income due to unemployment.

Other studies (e.g. Clark et al. 2009) find that the negative effect of unemployment on mental well-being itself is related to the regional unemployment rate. This result is

usually discussed in the context of social norms. The general findings state that being unemployed in a high unemployment rate region has a smaller negative effect on mental well-being as if the unemployed would live in a region with a low unemployment rate. In high unemployment rate regions being unemployed means to be conform to the social norm of unemployment. The results suggest that not deviating from the social norm lowers the psychological burden of unemployment. In contrast, Vatter (2012) found that subjective well-being in East Germany where unemployment rates are considerably higher than in West Germany is more affected when it comes to unemployment. He argues that lower job prospects in high unemployment rate regions increase the negative effect from unemployment. Clark et al. (2010) and Knabe and Rätzl (2011) provide empirical evidence for this relationship and show that current negative expectations about becoming re-employed in the future additionally reduce subjective well-being among the unemployed. These studies agree in that the negative effect from unemployment on mental well-being is heterogenous among individuals. Furthermore, the size of the negative effect depends on unemployment rates and future job prospects. But from the economic literature the mechanism how these components affect the perception of the employment status remains unclear.

De Witte (1999) provides a review of the psychological literature on the relationship of perceived job insecurity and psychological well-being. He summarizes from the literature that job insecurity reduces significantly the well-being in different psychological domains. He also analyzes the question how important job insecurity is compared to the effect of unemployment. His empirical findings suggest that the anticipation of unemployment has the same impact as unemployment on the psychological well-being. His results confirm a statement which Lazarus already made in 1966, that “the anticipation of harm can have effects as potent as experiencing the harm itself” (quoted by Roskies et al. (1993, p.619))

Dekker and Schaufeli (1995, p.58) state that in the psychological literature it has become apparent that the phase of job insecurity, in which termination is more or less anticipated, may very well be the most stressful aspect of the whole unemployment process. They compare two groups of employees in a large Australian public transport organization who at the same time faced uncertainties about whether or not they will become

unemployed due to organizational changes. They find that the psychological well-being of those who became unemployed in the next period improved compared to those who were still uncertain at this point. This result indicates that uncertainty about the future employment status not only affects mental well-being directly but also the perception of the unemployment status. Although one group of employees was finally made redundant they experienced an increase in their psychological well-being. They felt relieved from their uncertainty as they became unemployed according to their expectation.

Green et al. (2000) on the other hand analyzed which factors determine perceived job insecurity. They find that for the employed higher levels of unemployment rates increase perceived job insecurity, and higher levels and increases in unemployment rates also increase perceived difficulties of re-employment for the unemployed.

In this paper we bring together the several findings from the economic and psychology literature on unemployment and mental well-being and provide an explanation of the mechanism how unemployment rates and anticipation of unemployment affect the perception of unemployment based on economic theory. The theoretical foundation for the econometric analysis comes from models with reference-dependent preferences with endogenous reference points developed in the behavioral economics literature. These models formalize the effect of the anticipation of an event as well as the effect of a deviation of what an individual had expected as an outcome for this event.

Furthermore, our analysis differs from the previous studies in the sense that changes rather than levels of the employment status are analyzed and that not the influence of only current unemployment rates or job prospects on current mental well-being is measured but the effect of expectations and deviations from the expected employment status on the perception of unemployment. From a prospect theoretical point of view it seems more plausible that changes in the employment status rather than the absolute status influence mental well-being and that the valuation of unemployment depends on a certain reference point. Kahneman and Tversky (1979, p.277) state: "...the carriers of value are changes in wealth or welfare, rather than final states. This assumption is compatible with basic principles of perception and judgment. Our perceptual apparatus is attuned to the evaluation of changes or differences rather than to the evaluation of absolute magnitudes.

When we respond to attributes such as brightness, loudness, or temperature, the past and present context of experience defines an adaptation level, or reference point, and stimuli are perceived in relation to this reference point.” Therefore, differences in the perception of unemployment regarding mental well-being are probably not only the result of social norms that are somehow derived from the unemployment rates but from potential deviations of the individual employment status from what an individual had expected, i.e. his reference point. The literature on reference-dependence provides a discussion of the determination of reference points and mainly distinguishes exogenous and endogenous reference points. For our analysis the concept of endogenous expectation-based reference points proposed by Kőszegi and Rabin (2006; 2007; 2009) is applied. They propose that the individuals’s reference-point is determined by lagged expectations about outcomes rather than the status quo. Several studies recently addressed their research questions to the empirical evidence of reference points that are determined by expectations. Abeler et al. (2011) show in an real-effort laboratory experiment that labor supply is in line with the predictions of models with reference dependent preferences with reference points formed by expectations. Crawford and Meng (2011) re-analyze the labor supply of New York City cab drivers and find empirical evidence for reference-dependence preferences with expectation based reference points. Card and Dahl (2011) analyze violent behavior dependent on outcomes of football games. They find that for unexpected losses of the home team violence against partners significantly increases whereas expected losses of the football team have no significant effect on at-home violence.

In our context, unemployment rates serve as an information that determines reference points of the individuals and the magnitude of changes in mental well-being is related to the deviation of this reference point. More precisely, we assume that individuals observe relevant unemployment rates (e.g. industrial sector specific or regional unemployment rates) and that they use this information to build expectations about their future employment status. These current expectations serve as the reference point for the future employment status. Finally, the individuals compare the actual outcome of their employment status with their expected outcome. If the actual employment status deviates from the expected employment status we expect a stronger effect from this outcome com-

pared to the effect that arises when the actual employment status was already expected. More precisely for unemployment, we hypothesize that becoming unemployed is more has a more severe effect on mental well-being when unemployment hits the individual surprisingly rather than anticipated.

To test this hypothesis empirically it is essential to control for any unobserved individual level heterogeneity in mental well-being. As we focus on becoming or staying unemployed rather than being unemployed this leads naturally to a fixed effects estimator. We use the waves from 1998 to 2009 from the German Socio Economic Panel (SOEP) that provides all relevant information for our analysis.

In the next chapter we develop a simple theoretical model which motivates our empirical analysis. In chapter 3 we explain the regression model and the estimation strategy. Chapter 4 provides detailed information on the data set and variables used for the estimation. In chapter 5 we show and interpret the estimated effects. Finally, chapter 6 concludes.

2 Theoretical Framework

Theoretical models for reference-dependent preferences with endogenous reference points based on expectations from the behavioral economics literature deliver the theoretical background to our problem. These models support the idea that an individual is more affected by an outcome of an event that was not expected as if the same outcome was expected by the individual (see Section 1).

To motivate and structure the empirical analysis of reference-dependent effects of unemployment on mental well-being we borrow the formal structure of these theoretical models and substitute utility with the state of mental well-being. We can formalize the following theoretical model:

$$M_{it}(x_{it}, x_{it-1}, q_{it}, q_{it-1}) = u(x_{it} + x_{it-1}) + v(q_{it} + q_{it-1}) + \mu((1 - x_{it}) - q_{it-1}) \quad (1)$$

$$\begin{aligned} \text{with } x_{it} &= \begin{cases} 1 & \text{if } i \text{ unemployed in } t \\ 0 & \text{if } i \text{ employed in } t \end{cases} \\ \text{and } q_{it} &= \begin{cases} 1 & \text{if } i \text{ has positive expectations in } t \\ 0 & \text{if } i \text{ has negative expectations in } t \end{cases} \end{aligned}$$

Overall mental well-being $M_{it}(\cdot)$ for individual i at time t depends on the employment status x in t and $t - 1$, $u(\cdot)$; on expectations about the future employment status q in t and $t - 1$, $v(\cdot)$; and from a deviation of the current employment status in t from the expected employment status for t , $\mu(\cdot)$.

x_{it} describes the current employment status in t and takes the value 1 if the individual is unemployed in t and 0 if he is employed in t .

Expectations q are defined to be positive if an individual expects to be employed and to be negative if the individual expects to be unemployed in the future. For simplicity, we assume a binary outcome for expectations and q_{it} equals 1 for positive expectations in t about the employment status in $t + 1$, and q_{it} equals 0 for negative expectations in t about the employment status in $t + 1$.

As $M_{it}(\cdot)$ depends on employment status at two different points in time, t and $t - 1$, we can distinguish four different cases of employment histories:

- (1) i is employed in t and $t - 1$
- (2) i is unemployed in t and employed in $t - 1$
- (3) i is employed in t and unemployed in $t - 1$
- (4) i is unemployed in t and $t - 1$.

Table 1 summarizes the four different cases.

Moreover, $M_{it}(\cdot)$ depends on expectations in t and $t - 1$. Similarly to the unemployment histories we can distinguish 4 different case for the expectations histories:

Table 1: Employment histories

$x_{it-1} \backslash x_{it}$	0	1
0	(00)	(01)
1	(10)	(11)

- (i) negative expectations in $t - 1$ and t
- (ii) negative expectations in $t - 1$ and positive expectations in t
- (iii) positive expectations in $t - 1$ and negative expectations in t
- (iv) positive expectations in $t - 1$ and t

The four cases of expectations histories can appear in each of the four cases of employment histories. Therefore, we can finally distinguish 16 different types of individuals regarding their unemployment status and expectations over two periods in time. Table 2 shows the different combinations of expectations and employment histories.

Table 2: Expectations and employment histories

		q_{it-1}	0		1	
x_{it-1}	$q_{it} \backslash x_{it}$		0	1	0	1
0	0		(0000)	(0001)	(0010)	(0011)
	1		(0100)	(0101)	(0110)	(0111)
1	0		(1000)	(1001)	(1010)	(1011)
	1		(1100)	(1101)	(1110)	(1111)

Table 2 can be summarized in a compact employment-expectations matrix form with j rows and k columns.

$$Z = \begin{pmatrix} 0000 & 0001 & 0010 & 0011 \\ 0100 & 0101 & 0110 & 0111 \\ 1000 & 1001 & 1010 & 1011 \\ 1100 & 1101 & 1110 & 1111 \end{pmatrix}$$

Each element z_{jk} of the matrix contains the following information:

$$z_{jk} = (x_{it-1} \ x_{it} \ q_{it-1} \ q_{it}) \text{ with } j = 1, \dots, 4 \text{ and } k = 1, \dots, 4. \quad (2)$$

All individuals in the first row of the employment-expectations matrix were employed in $t - 1$ and t . All individuals in the second row were employed in $t - 1$ and unemployed in t . All individuals in the third row were unemployed in $t - 1$ and employed in t . All individuals in the last row were unemployed in $t - 1$ and t . All individuals in the first column had negative expectations in $t - 1$ and t . All individuals in the second column had negative expectations in $t - 1$ and positive expectations in t . All individuals in the third column had positive expectations in $t - 1$ and negative expectations in t . All individuals in the last column had positive expectations in $t - 1$ and t . For example, the individual denoted with (0000) was employed in $t - 1$ and t and had negative expectations in $t - 1$ and t , whereas the individual (0101) was employed in $t - 1$, unemployed in t , had negative expectations in $t - 1$ and positive expectations in t . Therefore, individuals (0110) and (0111) were employed in $t - 1$ but became unemployed in t although they had positive expectations about their employment status in $t - 1$. Thus, these individuals became unemployed unexpectedly. Respectively, individuals (1110) and (1111) remained unemployed unexpectedly.

From the current empirical literature on unemployment and mental well-being and the theoretical literature on reference-dependence the following two hypothesis on the relationship between unemployment and mental well-being can be derived:

- (i) In the case of becoming unemployed mental well-being deteriorates and in the case of becoming employed mental well-being increases.
- (ii) If an individual has expected his current employment status his mental well-being is less affected by the outcome of his actual employment status as if he would not have expected his current employment status. More precisely in the case of unemployment, if an individual expected to become unemployed then the negative

effect of becoming unemployed on mental well-being is less pronounced as if he would not have expected to become unemployed.

In the following Section the structure of the theoretical model formalized in this section is used as well as the consequential types of individuals to develop an econometric model that allows to test these hypothesis empirically.

3 Empirical Strategy

3.1 Empirical Model

In order to identify reference-dependent effects of unemployment on mental well-being empirically we translate Equation 1 into two different econometric models: first, a dummy variable model for all possible combinations of employment status and expectations in both periods, and second, model with pairwise interactions for employment status and expectations in both periods.

The dummy variable model follows straightforward from the theoretical model where 16 different cases of employment and expectations histories were distinguished. A dummy variable d is used for each of the cases. For simplicity we preliminarily abstract from any additional influencing factors as well as from unobserved heterogeneity (both will be introduced in the second regression model). We can write the following compact form of a linear regression model with y_{it} measuring mental well-being of individual i at time t :

$$y_{it} = \pi_0 + \sum_{j=1}^4 \sum_{k=1}^4 \pi_{jk} d_{jkit} - \pi_{11} d_{11it} + \epsilon_{it} \quad (3)$$

$$\text{with } d_{jkit} = \begin{cases} 1 & \text{if } (x_{it-1} \ x_{it} \ q_{it-1} \ q_{it}) = z_{jk} \\ 0 & \text{otherwise} \end{cases}$$

Expanding Equation 3 yields the following dummy variable model:

$$\begin{aligned}
y_{it} = & \pi_0 + \pi_{12}(0001)_{it} + \pi_{13}(0010)_{it} + \pi_{14}(0011)_{it} \\
& + \pi_{21}(0100)_{it} + \pi_{22}(0101)_{it} + \pi_{23}(0110)_{it} + \pi_{24}(0111)_{it} \\
& + \pi_{31}(1000)_{it} + \pi_{32}(1001)_{it} + \pi_{33}(1010)_{it} + \pi_{34}(1011)_{it} \\
& + \pi_{41}(1100)_{it} + \pi_{42}(1101)_{it} + \pi_{43}(1110)_{it} + \pi_{44}(1111)_{it} \\
& + \epsilon_{it}
\end{aligned} \tag{4}$$

Individual (0000) is arbitrarily chosen as the reference category. This model allows an immediate comparison of the mental well-being of different individuals. For example π_{13} reflects the difference in mental well-being of an individual who in $t - 1$ expected to stay employed in t and the reference individual who in $t - 1$ did not expect to stay employed in t , all else equal. In spite of its attractiveness for an easy comparison of individuals, this model does not allow for a non-ambiguous identification of reference-dependent effects of unemployment on mental well-being. Suppose we were interested in the effect of becoming unemployed unexpectedly. As already shown in the theoretical model this situation is given in two cases. In the dummy variable model the effect of unexpected unemployment is captured by the coefficients of all individuals who were employed in $t - 1$ and are unemployed in t and had positive expectations in $t - 1$. In this example this are the coefficients π_{23} and π_{24} (for individuals (0110) and (0111)). Both coefficients contain the effect from a deviation of the current employment status in t from the expected employment status for t . But, these two individuals differ in their current expectations in t about their future employment status in $t + 1$. This difference is also captured by the coefficients π_{23} and π_{24} . Thus, such a dummy variable model does not allow a unique identification of reference-dependent effects of unemployment. However, the structure of this model supports the later interpretation of the following econometric model with pairwise interactions of unemployment and expectations.

So far, we have not explicitly distinguished between different expectations about future employment status of the employed and the unemployed. The employed individuals build expectations about becoming unemployed or staying employed in the future. In contrast, the unemployed individuals build expectations about becoming re-employed or staying unemployed in the future. For the pairwise interacted model it will be differentiated

between the expectations of the employed and the unemployed (as it is also done in the data, see Section 4). For the employed individual i the expectation in t about his employment status in $t+1$ is denoted by \bar{q}_{it} . The expectation of an unemployed individual i in t about his employment status in $t+1$ is denoted by \underline{q}_{it} . The outcomes of both variables are defined analogous to the general expectation q_{it} in Equation 1:

$$\bar{q}_{it} = \begin{cases} 1 & \text{if the employed } i \text{ in } t \text{ expects to stay employed in } t+1 \\ 0 & \text{if the employed } i \text{ in } t \text{ expects to become unemployed in } t+1 \end{cases}$$

$$\underline{q}_{it} = \begin{cases} 1 & \text{if the unemployed } i \text{ in } t \text{ expects to become re-employed in } t+1 \\ 0 & \text{if the unemployed } i \text{ in } t \text{ expects to stay unemployed in } t+1 \end{cases}$$

Because, expectations \bar{q}_{it} and \underline{q}_{it} are mutually exclusive for individual i in t the distinction between expectations of the employed and unemployed was implicitly done before in the theoretical and the dummy variable model without loss of generality and in order to keep the notation easy.

With \bar{q}_{it} and \underline{q}_{it} the following pairwise interacted model that corresponds to Equations 1 and 3 can be obtained:

$$\begin{aligned} y_{ist} = & \beta_0 + \beta_1 x_{ist} + \beta_2 x_{ist-1} \\ & + \beta_3 \bar{q}_{ist} + \beta_4 \underline{q}_{ist} \\ & + \beta_5 \bar{q}_{ist-1} + \beta_6 \underline{q}_{ist-1} \\ & + \beta_7 (x_{ist} \times x_{ist-1}) \\ & + \beta_8 (\bar{q}_{ist} \times \bar{q}_{ist-1}) + \beta_9 (\bar{q}_{ist} \times \underline{q}_{ist-1}) \\ & + \beta_{10} (\underline{q}_{ist} \times \bar{q}_{ist-1}) + \beta_{11} (\underline{q}_{ist} \times \underline{q}_{ist-1}) \\ & + \beta_{12} (\bar{q}_{ist-1} \times x_{ist}) + \beta_{13} (\underline{q}_{ist-1} \times x_{ist}) \\ & + \beta_{14} (\bar{q}_{ist} \times x_{ist-1}) + \beta_{15} (\underline{q}_{ist} \times x_{ist-1}) \\ & + w_{ist} \beta + \alpha_i + \delta_s + \lambda_t + (\delta_s \times \lambda_t) + \varepsilon_{ist} \end{aligned} \tag{5}$$

y_{ist} measuring mental well-being of individual i in federal state s at time t . As mentioned earlier it is assumed that people use a certain unemployment rate to build expectations about their own employment status (see Section 1). In the empirical analysis we focus on unemployment rates at the federal state level.¹ In order to control for possible correlation between individuals at this level the federal state where each individual lives is additionally picked up by the subscript s .

As before, x_{ist} takes the value 1 if the individual i in federal state s is unemployed in t . \bar{q}_{ist} and \underline{q}_{ist} take the value 1 for positive expectations in t about the future employment status in $t + 1$ of the employed and the unemployed in federal state s , respectively.

To measure causal effects of unemployment on mental well-being it is necessary to control for any factors that influence mental well-being as well as unemployment. w_{ist} is a vector of control variables at the individual level. α_i captures all time-invariant unobserved individual heterogeneity. δ_s captures all time-invariant unobserved heterogeneity at the federal state level, and λ_t captures time fixed effects. The interaction of δ_s and λ_t controls for all federal state specific effects that vary over time. This includes for example unemployment rates at the federal state level but also more generally all time-variant unobserved heterogeneity between federal states. Modeling explicitly federal state specific time-variant heterogeneity captures all possible correlation between individuals in the same federal state. Instead of interactions also clustered standard errors at the level of federal states could have used to allow for correlation between individuals in the same federal states. But clustered standard errors at this level impose problems with those individuals who move between federal states and clustering in this case would need additional correction of degrees of freedom in the model. We also avoid the alternative to just exclude all individuals who moved between federal states as this would not only impose a general loss of information but could also lead to biased estimates due to selection if individuals who move between federal states systematically differ in their characteristics from individuals who not move. ε_{ist} is the usual idiosyncratic error term.

¹However, we have run the same regression models with data on the industrial sector level (2-digit NACE code) rather than on the federal state level. As the results barely change they are not shown in this paper but are available upon request from the author.

The coefficients β_1 to β_6 measure direct effects from current and past unemployment and current and past positive expectations. Thus, the linear coefficients β_1 to β_6 reflect the functions $u(\cdot)$ and $v(\cdot)$ in Equation 1 if we assume linearity for $u(\cdot)$ and $v(\cdot)$. The coefficients β_7 to β_{15} measure the effects of all possible pairwise interactions of the employment status and expectations in two subsequent periods of time. Although β_{12} and β_{13} are effects of unexpected unemployment, we cannot derive the function $\mu(\cdot)$ in Equation (1) straightforward from this model. Only certain linear combinations of coefficients allow the interpretation of effects as a reflection of $\mu(\cdot)$. When interpreting the estimated effects in Chapter 5 this will be explained in detail. Table 3 provides a detailed interpretation of those coefficients in the model that are related to the employment status and expectations.

In order to find the effects that uniquely identify reference-dependent effects of becoming or staying unemployed we can link the pairwise interacted model to the dummy variable model. Table 20 shows the relevant coefficients for each of the 16 cases. As shown in Section 2 the individuals (0110) and (0111) are those of interest as these became unemployed unexpectedly in t . The only difference between these two are their expectations in t . From Table 20 it can be seen that individual (0111) differs from individual (0110) in the coefficients β_4 and β_{10} . Both effects stem from the positive expectations that individual (0110) has in t in contrast to individual (0111). The coefficient that is unique for both individuals is β_{12} . This effect stems from the combination of positive expectations in $t - 1$, employment in $t - 1$ and unemployment in t , i.e. unexpected unemployment. Analogous, for the individuals (1110) and (1111) we find β_{13} to be the coefficient that identifies the effect of remaining unemployed unexpectedly as β_{13} stems from the combination of being unemployed in $t - 1$ and t but having positive expectations in $t - 1$. Therefore, for becoming or staying unemployed the coefficients β_{12} and β_{13} uniquely identify reference-dependent effects from unemployment on mental well-being, respectively. However, in order to have a meaningful comparison of individuals it will be necessary to compare certain linear combinations of coefficients. To test the hypothesis that an individual who became unemployed unexpectedly suffers more from becoming unemployed than an individual who already expected the unemployment the linear combination of β_5

Table 3: Variables, coefficients, and corresponding measured effects

Variable	Coefficient	Effect of
x_{ist}	β_1	current unemployment
x_{ist-1}	β_2	past unemployment
\bar{q}_{ist}	β_3	current positive expectations of the currently employed
\underline{q}_{ist}	β_4	current positive expectations of the currently unemployed
\bar{q}_{ist-1}	β_5	past positive expectations of the previously employed
\underline{q}_{ist-1}	β_6	previous positive expectations of the previously unemployed
$x_{ist} \times x_{ist-1}$	β_7	continued unemployment
$\bar{q}_{ist} \times \bar{q}_{ist-1}$	β_8	continued positive expectations of the continuously employed
$\bar{q}_{ist} \times \underline{q}_{ist-1}$	β_9	continued positive expectations of the previously unemployed and currently employed
$\underline{q}_{ist} \times \bar{q}_{ist-1}$	β_{10}	continued positive expectations of the previously employed and the currently unemployed
$\underline{q}_{ist} \times \underline{q}_{ist-1}$	β_{11}	continued positive expectations of the continuously unemployed
$\bar{q}_{ist-1} \times x_{ist}$	β_{12}	past positive expectations of the previously employed and current unemployment
$\underline{q}_{ist-1} \times x_{ist}$	β_{13}	past positive expectations of the previously unemployed and continued unemployment
$\bar{q}_{ist} \times x_{ist-1}$	β_{14}	current positive expectations of the currently employed and the previous unemployment
$\underline{q}_{ist} \times x_{ist-1}$	β_{15}	current positive expectations of the currently unemployed and the previous unemployment

and β_{12} (and additionally β_{10} in the case of positive expectations in t) is tested whether it is different from zero. The prediction is that this linear combination is negative, reflecting the additional negative effect that stems from the deviation of the expected employment status (i.e. 'employed in t ') from the actual employment status ('unemployed in t '). The detailed outline for the interpretation of the results is given in Chapter 5.2.

3.2 Estimation strategy

In order to identify a causal effect of unemployment on mental well-being it should be controlled for any heterogeneity that influences both mental well-being and unemployment. Panel data allow to account for any observed and unobserved determinants that are invariant over time or invariant over individuals or both. With interactions between federal states and time the model additionally controls for any federal state specific factors that vary over time (see Section 3.1). However, correlation over time within individuals that is not accounted for by any of the effects described can still be remaining. For example, unobserved factors at the individual level that evolve over time like life experience, perception of the relationship status, etc. that probably lead to a trend in mental well-being. This is reflected in an autocorrelated structure of the error term. Not accounting for such autocorrelation would lead to biased estimates of the standard errors of the coefficients and thus biased statistical tests. Therefore, heteroskedastic and autocorrelation consistent (HAC) standard errors by clustering at the individual level are estimated.

The dependent variable is subjective life satisfaction with outcomes on a scale from 0 (low) to 10 (high). Thus the dependent variable can be assumed to be cardinal as well as ordinal. Depending on the assumptions the estimation can be run with an linear estimator (e.g. ordinary least squares (OLS)) or an non-linear ordered latent response estimator (e.g. ordered probit or logit), respectively. Ferrer-i-Carbonell and Frijters (2004) provide an analysis of differences in estimated life satisfaction depending on the estimator. They show that using linear OLS and non-linear ordered response estimators essentially yield the same results for life satisfaction. They emphasize that controlling for time-invariant unobserved factors (individual fixed effects) matters to the estimates but not assumptions on cardinality or ordinality of life satisfaction. Therefore, we estimate mental well-being

with OLS and control for time-invariant unobserved heterogeneity.

3.2.1 Fixed and Random Effects Estimators

In general there are two different estimators that allow to control for unobserved individual specific heterogeneity, the fixed effects and the random effects estimator. The two estimators differ in the assumptions on the individual effects. The fixed effects estimator explicitly models time-invariant individual effects as a determinant of the dependent variable. By demeaning the data over time the fixed estimator controls for all constant individual heterogeneity but inherently removes variation from the covariates. Identification relies on variation within individuals. The random effects estimator is based on the assumption that the time-invariant individual effects are random and uncorrelated with all other explanatory variables and are modeled as part of a composed error term. Identification with the random effects estimator relies on variation within and between individuals. Therefore, if the assumption of randomness of the time-invariant individual effects holds the random effect estimator is more efficient than the fixed effects estimator.

Because the random and the fixed effects estimators differ in the source of identification one should be aware of the exact question that is to be answered in the analysis. Whereas the coefficient of unemployment estimated with the random effects estimator can be interpreted as the effect of being unemployed on mental well-being, the coefficient of unemployment estimated with the fixed estimator reflects the effect of becoming unemployed on mental well-being. In this paper we analyze the effect of unexpected changes in rather than levels of the employment status on mental well-being. This leads directly to the fixed effects estimator. Nevertheless, in our context the random effects estimator could still deliver reasonable interpretation of the coefficients when changes rather than levels of the variables are used in the empirical model. However, it is essential to check whether the crucial assumption about no correlation between unobserved heterogeneity and the observables holds. This can be tested by conducting a variable addition test (VAT), where the dependent variable is regressed on the regressor matrices X and \bar{X} (X demeaned over time by individuals).² The null hypothesis that the coefficients of \bar{X} are

²The standard Hausman test in this case is problematic as it does not allow to perform the test with

zero is tested with the classical F-Test.³

3.2.2 Estimation of Effects for Different Parts of the Population

The basic model is estimated for all individuals in the analysis data set (see Section 4). We are also interested whether certain groups in the population are affected differently from reference-dependent effects of unemployment on mental well-being and if the results from the basic estimation are robust for different parts of the population. It is focused on differences between genders and age groups. In order to keep the interpretation of results manageable, the basic pairwise interacted model (Equation 5) is estimated for various stratifications of the data rather than adding additional interactions terms for different groups in the population. Furthermore, in the model with individual fixed effects (see Section 3.2.1) only stratifying by gender allows the examination of gender specific differences. With individual fixed effects any time constant variables such as gender become zero when demeaning the data over time. Therefore a gender effect cannot be estimated with a fixed effects estimator.

4 Data

4.1 Sample

For the empirical analysis we use the waves from 1998 to 2009 from the German Socio-Economic Panel (SOEP). The SOEP started in 1984 with approximately 12 000 individuals in 6 000 households in West Germany and was extended to East Germany in 1990. After various sample refreshments the SOEP included more than 22 000 adult respondents in approximately 12 500 households in 2006 (Wagner et al., 2007). The data set contains information about the current employment status and expectations about the future employment status in each year. Unemployed individuals are asked for the reason

heteroskedastic and autocorrelation robust standard errors. Arellano (1993) developed a generalized VAT that is robust to autocorrelation and heteroscedasticity of arbitrary forms.

³See Baltagi (1998) and Arellano (1993) for further details and properties of this test.

of their job termination. The SOEP also provides various measures for mental well-being (see below for further details) and socio-economic characteristics of the individuals.

In order to analyze the relevant part of the population we reduce the observations for the final analysis data set. Only those individuals are kept who are part of the economically active population. Therefore, the sample is restricted to individuals who are of age 30 to 55. The lower bound is to avoid the part of the population that is most probably still under education, or in a pecuniary and instable job situation. The upper bound is to avoid those individuals who already could face special incentives for job termination due to proximity to retirement, e.g. special regulations for early retirement. Among the employed individuals only those who are in full-time employment enter the sample. The restrictions imposed on the unemployed in the sample are in order to make these individuals most comparable to the employed. Therefore, only legally registered unemployed individuals who intend immediate full-time re-employment are kept. As the dependent variable in the model is a measure of mental well-being most probably reverse causality between mental well-being and unemployment would appear. People with mental health problems plausibly have a higher probability of becoming unemployed due to less productivity. Without further restrictions the estimation could suffer from an endogeneity problem. Following Schmitz (2011) we concentrate on those individuals with exogenous entries into unemployment due to plant closures to minimize the potential bias in the estimation of the effect of unemployment on mental well-being due to the endogeneity of unemployment.⁴

As a proxy for mental well-being we use life satisfaction that is self rated on a scale of 0 (low) to 10 (high). The SOEP would also allow to analyze the relationship between unemployment and mental health rather than mental well-being as it provides a measure for mental health, the Mental Component Summary Scale (MCS)⁵. But MCS is provided

⁴Schmitz (2011) shows that the general finding that unemployment has a negative effect on health is likely to be the result of biased estimates and does not reflect a causal relationship. With only plant closures as exogenous entries into unemployment he does not find any effect of unemployment on various health measures.

⁵MCS is an scale score that is calculated using explorative factor analysis with various self-reported measures of mental health in the SOEP (see Andersen et al. (2007) for further information on the

only every two years. As our model requires observations of two consecutive periods we loose too many observations with MCS. Therefore we concentrate on life satisfaction as a proxy for mental well-being. Table 4 shows correlation coefficients and p-values of life satisfaction and MCS with particular self-reported mental-health measures provided biannually in the SOEP.

Table 4: Life satisfaction and mental health measures

Variables	Life Satisfaction		MCS	
MCS	0.3940	(0.000)		
pressed	-0.1937	(0.000)	-0.3567	(0.000)
melancholy	-0.4158	(0.000)	-0.6645	(0.000)
balanced	0.3644	(0.000)	0.6011	(0.000)
energy	0.3534	(0.000)	0.5376	(0.000)
acclessmental	-0.3717	(0.000)	-0.6737	(0.000)
lesscaremental	-0.3325	(0.000)	-0.6438	(0.000)
<i>N</i>	23485			

Note: p-values in parenthesis

Not surprisingly the correlations between life satisfaction and certain mental-health measures are less pronounced than the correlations between MCS and the same mental-health measures as MCS is calculated on the basis of these variables. However, all correlations show the same sign as with MCS and are highly statistical significant. The strength of correlation varies more between MCS and mental-health measures than for life satisfaction. The correlation of life satisfaction and MCS is 0.39 and highly statistical significant. Therefore, life satisfaction could be interpreted as a proxy for mental health as well as for mental well-being. Nevertheless, we will interpret general life satisfaction as a measure of mental well-being as this interpretation seems to be more adequate in the context of utility which is the dependent variable in the theoretical models for reference-dependence.

algorithm).

4.2 Variables and Descriptive Statistics

Table 5 reports summary statistics for the key variables of the analysis. The total number of observations in the analysis sample is 62135. The share of legally registered unemployed individuals is only 0.74%. Rather than reflecting the true population unemployment rate in Germany this low share of unemployed is caused by the selection process of observations described above. The requirement of only exogenous entries into unemployment is very restrictive and a high fracture of unemployed individuals does not enter the sample.

Table 5: Summary statistics of life satisfaction, employment status, and expectations

Employed												Unemployed																							
$N = 61\,687$												$N = 457$																							
99.26 %												0.74 %																							
$N = 62\,135$																																			
Employment Expectations																																			
very				somewhat				not at all				impossible				difficult				easy															
15.4 %				44.3 %				40.3 %				21.2 %				74.3 %				4.5 %															
Life Satisfaction																																			
low						\emptyset						high						low						\emptyset						high					
0	1	2	3	4	5	6	7	8	9	10		0	1	2	3	4	5	6	7	8	9	10		0	1	2	3	4	5	6	7	8	9	10	
0.2	0.2	0.9	2.1	3.1	10.3	11.6	25.2	32.8	10.8	2.8		1.1	2.0	5.0	9.6	9.4	25.2	14.2	16.4	13.6	2.6	0.9		1.1	2.0	5.0	9.6	9.4	25.2	14.2	16.4	13.6	2.6	0.9	

Expectations about the future employment status are different for the employed and the unemployed (see also Section 3.1). In the SOEP questionnaires the employed individuals are asked about their concerns about their job security and can choose between three possible answers: very concerned, somewhat concerned, and not concerned at all. The unemployed are asked about their perceived difficulties to find an appropriate position and can choose between the categories: easy, difficult, and almost impossible. Comparing the distribution of answers over the three ordered categories of expectations very distinctive patterns for the employed and unemployed appear. About 40% of the employed are not concerned at all about their job security. But only 4.5% of the unemployed believe that it will be easy for them to find a new job. 15.4% of the employed are very concerned about

their job security and 21.2% of the unemployed expect that it will be almost impossible to find a new job. Whereas 44.4% of the employed report to be somewhat concerned about their job security, 74.3% of the unemployed expect to have difficulties to find a new job. The descriptive statistics suggest that the unemployed tend to more pessimistic expectations about their employment future than the employed. In order to keep the interpretation of the estimated effects manageable we collapse the expectations into binary variables. Therefore, according to the theoretical model in Section 2 we will interpret the effect of positive expectations with reference to negative expectations. The response categories deliver a natural cut-off between negative and positive expectations (only the categories 'not at all' and 'easy' have a non-negative comprehension). Therefore, a dummy variable for positive expectations for the employed is defined that takes the value 1 for individuals who are 'not concerned at all' about their job security (corresponding to \bar{q}_{it} in Section 3.1). The choice of the cut-off between categories for the expectations of the unemployed is unfortunately not that clear-cut. The category 'easy' would be the natural outcome to define positive expectation. However, it cannot solely used because of the low share of respondents in this category. With only 4.5% of the unemployed in this category there would not be enough variation in the binary variable for positive expectations. Therefore, the cut-off is chosen between 'impossible' and 'difficult' and a dummy variable for positive expectations for the unemployed is defined that takes the value 1 if for individuals who expect that finding a new job will be 'easy' or 'difficult' but not 'impossible' (corresponding to \underline{q}_{it} in Section 3.1).

The dependent variable in the model is life satisfaction as a proxy for mental well-being. Individuals are asked to rate their overall life satisfaction on a scale from 0 (low) to 10 (high). The distribution of answers on this scale again is different for the employed and the unemployed. Whereas about 90% of the employed rate their life satisfaction between 5 and 9 with a peak in 8, the variance of life satisfaction is higher for the unemployed. The mean life satisfaction for the employed is 7.1 and for the unemployed 5.5 (see also table 6). The standard errors for life satisfaction for the employed and unemployed are 1.59 and 2, respectively. Without controlling for any additional factors the average difference in life satisfaction between the employed and the unemployed is about 1.6 points.

Table 6 additionally reports summary statistics for the control variables by employment status. We control for age, years of education, marital status (binary), number of children living in the same household, monthly net income, citizenship (binary), private health insurance (binary), blue-collar employment (binary), and self assessed health (scale from 1 (low) to 5 (high)). The employed and unemployed are on average similar in the control factors except net income and private health insurance. The fact that some unemployed have a positive net income at all (mean is 1.77 Euro per month) is because to some threshold the unemployed are allowed to earn some money without having affected their legal unemployment status and their unemployment benefits. Only 1% of the unemployed are privately insured compared to 13% of the employed. This difference can be explained by the German institutions for health insurance. In general only high income earners, self-employed, and civil servants are allowed to opt out from the public health insurance. When becoming registered as unemployed the privately insured typically have to switch back into the public system. However, there are some exceptions from this and under certain circumstances the unemployed are allowed to stay in the private system (mainly on their own expenses).

Table 6: Summary statistics for controls and dependent variable

Variable	All		Unemployed		Employed	
	Mean	SD	Mean	SD	Mean	SD
Life Satisfaction	7.05	1.60	5.46	2.00	7.06	1.59
Age	42.72	7.00	42.45	6.96	42.72	7.00
Years of Education	12.59	2.70	11.68	2.43	12.60	2.70
Married	0.71	0.45	0.58	0.49	0.71	0.45
Number of Children in Household	0.79	0.97	0.81	1.02	0.79	0.97
Net Income	1675.50	992.77	1.77	37.90	1687.90	985.89
Foreign	0.07	0.26	0.08	0.27	0.07	0.26
Private Insurance	0.13	0.34	0.01	0.11	0.13	0.34
Blue Collar	0.33	0.47	-	-	0.33	0.47
Self Assessed Health	3.55	0.81	3.48	0.91	3.55	0.81
<i>N</i>	62135		457		61678	

5 Results

5.1 Variable Addition Test for Unobserved Heterogeneity

In Section 3.2.1 the importance of testing for correlation between unobserved heterogeneity and the observed variables included in the model in order to decide whether the random effects estimator is applicable to our analysis was emphasized. We performed a VAT (see Section 3.2.1) following Arellano (1993). The usual F-Test rejects the joint null hypothesis that all coefficients of the demeaned explanatory variables are zero at the 0% significance level for all models (including all stratifications). Thus it is rejected, that none of the unobserved time-invariant heterogeneity captured by means over time is uncorrelated with the observed explanatory variables. So the random effects estimator is not applicable in our case as its crucial assumption of independence of the unobserved heterogeneity is rejected. Therefore, we rely the interpretation of the estimated effects of on the results from the fixed effects estimation.

5.2 Interpretation Strategy of the Results

For the interpretation of effects the results are examined in three steps following the structure of the two empirical models that were introduced in Section 3.1.

First, individuals that are employed and unemployed in t each with the same expectations history and the same employment status in $t - 1$ are compared pairwise. In particular, we compare the following pairs that were employed in $t - 1$: (0000) and (0100), (0001) and (0101), (0010) and (0110), and (0011) and (0111). We compare the following pairs that were unemployed in $t - 1$: (1000) and (1100), (1001) and (1101), (1010) and (1110), and (1011) and (1111). Applying hypothesis tests for multiple coefficients and calculating linear combinations of coefficients this kind of comparison allows to analyze whether or not comparable employed and unemployed individuals differ significantly in their mental well-being at all and to quantify the magnitude of such a difference.

In the second step, individuals that were employed in $t - 1$ and became unemployed in t but with different expectations regarding their employment status in t are compared. In particular, we compare individuals (0101) and (0111), and (0100) and (0101). It is

tested whether individuals who became unemployed unexpectedly differ from individuals who expected their unemployment.

Finally, the coefficient that uniquely measures the effect that originates from the unexpectedness of unemployment on mental well-being is interpreted in order to quantify the reference-dependent effect of unemployment on mental well-being.

The interpretation of the results follows the same three step structure for all stratifications.

5.3 Results from Fixed Effects Estimation for All Individuals

Table 7 reports the estimated OLS coefficients for the pairwise interacted fixed effects model applied to the whole sample. The estimates correspond to the coefficients in Equation 5 of the theoretical regression model introduced in Section 3.2.

5.3.1 Differences between employed and unemployed

As explained above we first concentrate on the difference in mental well-being between the employed and unemployed. Table 8 shows the results of calculated and F-tested linear combinations of estimated coefficients that reflect the differences in mental well-being between comparable pairs of employed and unemployed individuals. The second and the third columns report comparisons of currently employed and unemployed. Whereas in the second column both individuals were employed in $t - 1$ the individuals compared in the third column were both unemployed in $t - 1$. In the rows the pairs of currently employed and unemployed are separated by their histories of expectation.

The first cell shows the difference in mental well-being of currently employed and currently unemployed individuals where both individuals were employed in $t - 1$ and both had negative expectations in $t - 1$ and in t and all else equal. The mental well-being of this pair differs in the coefficient β_1 and is on average 0.3032 points lower for these unemployed than the mental well-being of the compared employed. The null hypothesis that β_1 equals zero cannot be rejected at a significance level lower than 19.02%. Therefore, we do not find a significant difference in the mental well-being of employed and unemployed with currently negative expectations when both were employed and had negative expectations

in the past period.

Comparing currently employed and unemployed with negative expectations in both periods but unemployment in $t - 1$ the linear combination of β_1 and β_7 is not statistically significant different from zero just at the 10% significance level. Thus, independent from the past employment status we find no statistically significant difference in the mental well-being of employed and unemployed if negative expectations are present in t and $t - 1$. The effects of negative expectations in two consecutive periods seem to dominate any difference in mental well-being between employed and unemployed individuals that stems from the difference in the employment status.

The mental well-being of currently employed and unemployed differs statistically highly significant for all other combinations of expectations and employment histories. The highest difference in mental well-being appears between those employed and unemployed who had negative expectations in $t - 1$ but positive expectations in t (between (1001) and (1101)). In this case we observe the difference in mental well-being of an individual that became employed unexpectedly in t (a positive deviation from the reference point) and adjusted expectations in t and an individual that remained unemployed expectedly (no deviation from the reference-point) and also with positive expectations in t . This result can be seen as a first empirical hint to reference-dependence in the context of employment and unemployment. Also, the average difference of 1.91 and 1.72 points in mental well-being of the employed and unemployed with positive expectations in $t - 1$ and negative expectations in t and past employment and unemployment respectively is not only statistically significant but substantial. In both cases we observe individuals who became unemployed unexpectedly and adjusted their expectations in t downwards. Thus, the comparison of employed and unemployed individuals already shows evidence for reference-dependent effects of the employment status on mental well-being as the biggest differences in mental-well being can be found for those cases where a change in the employment status was unexpected.

5.3.2 Differences between expected and unexpected unemployment and the reference-dependent effect

In the following we concentrate on those individuals that became unemployed unexpectedly. Becoming unemployed unexpectedly requires employment and positive expectations in $t - 1$ and unemployment in t . Therefore, the individuals of interest are (0110) and (0111). Both were employed in $t - 1$, are unemployed in t , and had positive expectations in $t - 1$. The only difference in both individuals lies in their expectations in t . Individual (0110) has in t negative expectations about becoming re-employed in $t + 1$ whereas individual (0111) has in t positive expectations. As we are interested in the effect of unexpected unemployment on mental well-being we compare the two types of unexpected unemployed to those unemployed individuals who expected their unemployment but have the same expectations in t and all else equal. In this sense the compared individuals have the same employment histories and the same expectations in t but differ in their expectations in $t - 1$. This makes two comparable pairs: individuals (0100) versus (0110), and (0101) versus (0111). Both pairs were employed in t and unemployed in $t - 1$. Within both pairs the individuals differ in their expectations in $t - 1$ but agree in their expectations in t . Between pairs the difference lies in their expectations in t , where the first pair has negative expectations and the latter pair positive expectations in t .

Table 9 shows the results of calculated and F-tested linear combinations of estimated coefficients that reflect the differences in mental well-being between comparable pairs of individuals who became expectedly and unexpectedly unemployed.

The first cell shows the estimated average difference in mental well-being between unexpected and expected unemployed with negative expectations in t for both. The difference in mental well-being between these two individuals is reflected by the linear combination of β_5 and β_{12} . The estimated difference in life satisfaction is 1.54 points. The F-test rejects the null hypothesis that the linear combination of β_5 and β_{12} equals zero at an acceptable 5.75% significance level. Thus, an individual that did not expect to become unemployed has on average a by 1.5 points lower life satisfaction compared to an individual who expected his unemployment, all else equal. This applies for unexpected unemployment, when expectations about future employment are adjusted downwards in the

period of becoming unemployed. In contrast, we find no statistically significant difference in mental well-being between unexpected and expected unemployed when expectations remain and become positive in t , respectively. The distinguishing linear combination of coefficients in this case is $\beta_5 + \beta_{10} + \beta_{12}$. The estimated coefficient of this linear combination is -0.04 points in life satisfaction and is not statistically significant different from zero. Thus, individuals who became unemployed unexpectedly but who have still positive expectations about their future employment status are not different from those unemployed who expected to become unemployed but also have positive expectations about their future employment. Thus, depending on the expectations in t we find a reference-dependent effect of unemployment on mental well-being. Whereas individuals with current positive expectations seem not to be affected by the fact that their unemployment was not expected, we find for individuals who are pessimistic about their future employment status a clear negative effect that stems from the unexpectedness of their unemployment. A detailed look at the estimated coefficients in the particular linear combinations reveals the mechanism of this difference.

Again, Table 7 shows in particular the estimated coefficients that contribute to the calculation of the linear combinations above. First of all, the coefficient of the variable $(\bar{q}_{ist-1})(x_{ist})$, β_{12} , is the one which uniquely measures the reference-dependent effect of becoming unemployed unexpectedly. For both types of individuals who became unexpectedly unemployed this coefficient is part of the linear combinations of coefficients that distinguish them from the expectedly unemployed. The estimate is -1.6 and statistically significant at the 5% level. For both types of unexpectedly unemployed individuals this result shows an average drop in life satisfaction of almost 2 points. Again, this negative effect only stems from the unexpectedness of their unemployment. However, both types benefit from their positive expectations in $t - 1$. This effect is reflected in the coefficient of the variable \bar{q}_{ist-1} , β_5 . The estimate of this coefficient is 0.07 and statistically significant at a lower than 1% level. However, compared to individuals who expected their unemployment, all else equal, this positive effect is not able to outweigh the negative effect from the unexpected unemployment. Therefore, the results suggest that the unexpected incidence of unemployment worsens the situation for the unemployed. Focusing

on those unexpectedly unemployed with ongoing positive expectations the coefficient of the variable $(q_{ist})(\bar{q}_{ist-1})$, β_{10} , is of additional relevance. This coefficient captures the effect of continued positive expectations in the case of becoming unemployed in t . The estimate is 1.5 and statistically significant at the 10% level. In absolute values the estimate of β_{10} is close to the estimate of β_{12} . The fact that these individuals in spite of their unexpected unemployment go on with positive expectations makes them statistically not distinguishable to those individuals who expected their unemployment. The positive effect from ongoing positive expectations outweighs the negative effect from unexpected unemployment.

In summary these results show a general reference-dependent negative effect for all individuals that became unemployed unexpectedly. This effect stems from the unexpectedness of unemployment, i.e. a negative deviation from the reference point. Individuals who have negative expectations about their job future after they became unemployed unexpectedly, i.e. individuals who adjusted their expectations downwards after becoming unemployed unexpectedly directly suffer from the negative deviation of their employment status from their reference point. Their positive expectations in the period prior to their unemployment cannot outweigh the negative effect from the unexpected unemployment. In contrast, individuals who became unemployed unexpectedly but with unaffected positive expectations about their future employment are statistically not different from those who became unemployed expectedly. This similarity is owed to the fact that in this case the positive effect from ongoing positive expectations outweighs the negative effect from unexpected unemployment.

5.4 Results from Fixed Effects Estimation by Age Groups

In order to estimate different slopes of the regression line for different ages the data set is stratified in two age groups (similar to using interaction terms). The number of only two sub-samples is mainly driven by the limited number of observed unemployed individuals. The first sub-sample includes individuals of age 30 to 40 (24 731 observations) and the second sub-sample includes individuals of age 41 to 55 (37 404 observations). Table 21 shows the distributions of life satisfaction and expectations over age years by employment

status. Average life satisfaction in the older age groups (6.99 and 5.31 for the employed and unemployed, respectively) is slightly lower than in the younger group (7.16 and 5.18 for the employed and unemployed, respectively). However, the average share of unemployed with positive expectations in the older age group is about 12% points lower than in the younger age group (83.4% and 71.7%). There is no such a clear difference in average expectations between younger and older employed individuals (40.7% and 40.1%).

The estimated coefficients for the younger age group mainly confirm the findings from the basic estimation, see Table 15. Table 16 reports the results for the estimated differences in life satisfaction between employed and unemployed young individuals. Again, the highest differences in life satisfaction between employed and unemployed individuals is found for unexpected outcomes in the employment status with adjusted expectations in the next period. The estimated difference between individuals who became unexpectedly employed and individuals who expectedly remained unemployed with an upward adjustment of expectations, (1010) and (1110), is 3.87 points in life satisfaction (0.2% significance level). Also, individuals who became unexpectedly unemployed followed by downward adjusted expectations have a 2.8 points lower life satisfaction than comparable employed individuals (0.0% significance level). In both cases the difference in life satisfaction exceeds the overall difference in life satisfaction between employed and unemployed individuals, see Section 4.2.

For the older age group the results suggest only minor differences in life satisfaction between employed and unemployed individuals, see Table 17. Moreover, the unexpectedness of the employment status outcome in cases with adjusted expectations in the next period seems not to play a role. Only in the case with unexpected ongoing unemployment and downward adjusted expectations ((1010) versus (1110)) the estimated difference in life satisfaction of 1.88 points is significant at the 0.0% level.

These findings for both age groups are also reflected in the estimated differences in life satisfaction between expected and unexpected unemployed individuals, see Table 18 and Table 19. For the younger age group the reference-dependent effect (β_{12}) is -1.9 and highly significant. This effect is only slightly lowered by the highly significant effect of previous positive expectations (β_5), 0.1. Thus, the overall reduction in life satisfaction

that occurs because the unemployment was not expected is estimated with 1.8 points (0.1% significance level). In the case where positive expectations are not affected by the unemployment a significant reference-dependent effect does not appear.

As expected from the comparison of life satisfaction levels between employed and unemployed individuals in the older age group the results suggest no empirical evidence for reference-dependent effects of unemployment for this part of the population.

A general higher fluctuation in the job market for younger individuals in the data set could be supposed as a possible explanation for this result. Table 23 shows the numbers of observations for all appearing counts of total unemployment periods per individual. The distribution of total counts is almost the same for the younger and older age group. Thus, a higher volatility for younger individuals between employment and unemployment periods seems not to be the reason for our findings.

Another explanation could be that younger individuals tend to be less risk averse than older individuals. Therefore, they might choose jobs with general lower job security such as in young and developing startup companies with a higher probability than more risk averse older individuals. However, we run the same regression with industrial fixed effects instead of federal state fixed effects and find similar results.

We tend towards the level of expertness on the job market as the most plausible explanation for the difference between younger and older individuals. Whereas older individuals might be more experienced in the evaluation of information regarding their future employment status, younger individuals seem to be less able to anticipate potential unemployment. The difference in the ability of foreseeing unemployment between younger and older individuals is supported by the data, see Table 22. 19.1% of the younger individuals who became unemployed did not expect their unemployment, whereas only 5.2% of the older age group became unemployed without expecting it. Not such a clear but similar pattern can be found for those individuals who stayed unemployed. 75.8% of the younger unemployed in $t-1$ who stayed unemployed in t had positive expectations for t whereas the share amongst the older unemployed is 70%. These numbers suggest that too few individuals in the older age class did not expect to become unemployed to show a statistically significant reference-dependent effect of unemployment on mental well-being

among this group.

5.5 Results from Fixed Effects Estimation by Gender

The estimated coefficients for the stratified data by gender mainly confirm the findings from the basic estimation.

For men we find a statistically significant lower life satisfaction by 1.3 points on average for unemployed even with negative expectations in $t - 1$ and t and unemployment in $t - 1$ for both, see Table 11. However, there is no significant difference between employed and unemployed males when both were employed in $t - 1$ and unemployment was not expected and expectations adjusted downwards in t ((0010) versus (0110)). This result is also reflected in Table 13. Whereas the results of the basic estimation suggest a statistically significant difference in life satisfaction of expectedly and unexpectedly unemployed individuals with downward adjusted expectations in t this is not the case for males. The linear combination of β_5 and β_{12} shows a lower life satisfaction by 1.3 points for unexpectedly unemployed but the difference is not statistically significant. For a deeper insight we estimated a further stratification for males by age and find that only for males of age 41 to 55 no reference-dependent effect appears. For males of age 30 to 40 we find a drop in life satisfaction by 2.3 points on average (0.3% significance level) caused by the unexpectedness of unemployment when expectations are adjusted downwards. Again, this result reflects the differences between age groups as discussed in Section 5.4. Interestingly, for younger men when expectations are adjusted upwards or remain positive after becoming unemployed the effect of ongoing expectations not only outweighs the negative effect from unexpected unemployment but even exceeds it ($\beta_{10} = 3.27$ at 0.1% significance level).

For females we find similar results as in the basic estimation as well as in the comparison of employed and unemployed (see Table 12) as in the comparison of expected and unexpected unemployment (see Table 13). In the case of downward adjusted expectations after becoming unemployed unexpectedly the average reference-dependent effect is -2.11 points in life satisfaction. With no adjustment of expectations we find no reference-dependent effect for females. The results of the regression for the further stratified female sub-sample into age groups suggest no differences between older and younger women

regarding reference-dependent effects of unemployment on mental well-being.

Summarizing this subsection, we find empirical evidence for reference-dependent effects of unemployment on mental well-being for women and young men. Only for older men the results suggest no evidence for reference-dependence in the context of unemployment.

6 Conclusion

Our empirical results show that mental well-being of individuals who expected to become unemployed is less affected from becoming unemployed as if the unemployment was not expected previously. We find that current and past expectations about the future employment status have an important impact not only directly on mental well-being but also on the perception of the employment status. Our results are derived from the estimation of an econometric model which follows the structure of theoretical models with reference dependent preferences and endogenous reference points that are determined by lagged expectations. We assumed that unemployment rates are used as an information to build expectations about the future employment status and lagged expectations represent the reference point. We developed the hypothesis that depending on expectations (i.e. the reference point) becoming unemployed affects the individuals differently.

The contribution of our study is twofold. First, we add to the literature on unemployment and mental well-being where the mechanism of how unemployment rates and expectations affect the perception of unemployment remained unclear so far. Whereas in this strand of literature only current expectations about the future are taken into account we show that past expectations play an important role in the perception of unemployment. We find that previously expecting unemployment attenuates the negative effect from becoming unemployed. It seems important to give individuals sufficient notice of their unemployment so that they are able to anticipate their unemployment and probably adapt to this situation. On the other hand it can be important to re-employment programs to focus on individuals who became unemployed unexpectedly in particular as the higher drop in mental well-being may involve a higher risk of developing serious mental illnesses. This in turn can reduce the chances of re-employment. Our results show that

positive expectations about re-employment even in the case of unexpected unemployment are able to keep up mental well-being at the level as if the unemployment was expected.

Second, our finding that unexpected unemployment has a stronger negative impact on mental well-being than expected unemployment supports theoretical models with reference dependent preferences and endogenous reference point formation with empirical evidence. Therefore, we also contribute to the literature on the importance of reference points (DellaVigna (2009) for an overview). Our results suggest that lagged expectations about the future employment status indeed serve as reference point and that the size of the effect of unemployment on mental well-being reflects a deviation from an individual reference point rather than the final state of unemployment.

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Appendix

Table 7: Fixed effects estimates for life satisfaction – all individuals

Variable	Coefficient	HAC SE
x_{ist}	β_1 -0.3032	0.2314
x_{ist-1}	β_2 -0.1258	0.1371
\bar{q}_{ist}	β_3 0.2147***	0.0195
\underline{q}_{ist}	β_4 -0.4892*	0.2593
\bar{q}_{ist-1}	β_5 0.0642***	0.0190
\underline{q}_{ist-1}	β_6 0.0399	0.1459
$x_{ist} \times x_{ist-1}$	β_7 -0.6005	0.5949
$\bar{q}_{ist} \times \bar{q}_{ist-1}$	β_8 -0.0198	0.0264
$\bar{q}_{ist} \times \underline{q}_{ist-1}$	β_9 -0.6130*	0.3709
$\underline{q}_{ist} \times \bar{q}_{ist-1}$	β_{10} 1.4960*	0.8750
$\underline{q}_{ist} \times \underline{q}_{ist-1}$	β_{11} 1.2091	0.8602
$\bar{q}_{ist-1} \times x_{ist}$	β_{12} -1.6021**	0.8098
$\underline{q}_{ist-1} \times x_{ist}$	β_{13} -0.8131	0.6450
$\bar{q}_{ist} \times x_{ist-1}$	β_{14} 0.6851*	0.3513
$\underline{q}_{ist} \times x_{ist-1}$	β_{15} 0.0329	0.7730
Age	-0.0298*	0.0137
Years of Education	-0.0185	0.0174
Married	0.1265**	0.0361
Children in household	0.0189	0.0140
Net Income	0.0001**	0.0000
Foreign	0.1573	0.1233
Private insurance	0.0440	0.0463
Blue collar	-0.0359	0.0311
Self assessed health	0.4582**	0.0107
Constant	6.6014**	0.6638
α_i	yes	
δ_s	yes	
λ_t	yes	
$(\delta_s)(\lambda_t)$	yes	
N	62135	

Note: *p<0.1, **p<0.05, ***p<0.01

Table 8: Employed versus unemployed – all individuals

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/0) vs (0/1)	(1/0) vs (1/1)
(0/0)	$H_0: \beta_1 = 0$ p-value = 0.1902 $\beta_1 = -0.3032$	$H_0: \beta_1 + \beta_7 = 0$ p-value = 0.1005 $\beta_1 + \beta_7 = -0.9037$
(0/1)	$H_0: \beta_3 = \beta_1 + \beta_4$ p-value = 0.0000 $(\beta_1 + \beta_4) - (\beta_3) = -1.0071$	$H_0: \beta_3 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{15}$ p-value = 0.0021 $(\beta_1 + \beta_4 + \beta_7 + \beta_{15})$ $-(\beta_3 + \beta_{14}) = -2.2599$
(1/0)	$H_0: \beta_1 + \beta_{12} = 0$ p-value = 0.0142 $\beta_1 + \beta_{12} = -1.9053$	$H_0: \beta_1 + \beta_7 + \beta_{13} = 0$ p-value = 0.0000 $\beta_1 + \beta_7 + \beta_{13} = -1.7168$
(1/1)	$H_0: \beta_3 + \beta_8 = \beta_1 + \beta_4 + \beta_{10} + \beta_{12}$ p-value = 0.0005 $(\beta_1 + \beta_4 + \beta_{10} + \beta_{12}) - (\beta_3 + \beta_8) = -1.0934$	$H_0: \beta_3 + \beta_9 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15}$ p-value = 0.0000 $(\beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15})$ $-(\beta_3 + \beta_9 + \beta_{14}) = -1.2509$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 9: Expected versus unexpected unemployment – all individuals

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/1)
(0/0) vs (1/0)	$H_0: \beta_5 + \beta_{12} = 0$ p-value = 0.0575 $(\beta_5 + \beta_{12}) = -1.5379$
(0/1) vs (1/1)	$H_0: \beta_5 + \beta_{10} + \beta_{12} = 0$ p-value = 0.9008 $(\beta_5 + \beta_{10} + \beta_{12}) = -0.0419$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 10: Fixed effects estimates for life satisfaction – by gender

		Male		Female	
Variable		Coefficient	HAC SE	Coefficient	HAC SE
x_{ist}	β_1	-0.1930	0.3101	-0.5517*	0.3116
x_{ist-1}	β_2	-0.0874	0.1951	-0.1488	0.1895
\bar{q}_{ist}	β_3	0.2008***	0.0259	0.2332***	0.0296
\underline{q}_{ist}	β_4	-0.8133**	0.3451	0.1529	0.3543
\bar{q}_{ist-1}	β_5	0.0541**	0.0240	0.0755**	0.0306
\underline{q}_{ist-1}	β_6	-0.0079	0.2042	0.0880	0.2070
$x_{ist} \times x_{ist-1}$	β_7	-1.0904*	0.6440	0.0749	0.9737
$\bar{q}_{ist} \times \bar{q}_{ist-1}$	β_8	0.0130	0.0339	-0.0615	0.0419
$\bar{q}_{ist} \times \underline{q}_{ist-1}$	β_9	-1.3243***	0.4752	-0.1012	0.5319
$\underline{q}_{ist} \times \bar{q}_{ist-1}$	β_{10}	1.4514	1.1996	1.9160**	0.8826
$\underline{q}_{ist} \times \underline{q}_{ist-1}$	β_{11}	1.0028	0.9886	1.6252	1.2343
$\bar{q}_{ist-1} \times x_{ist}$	β_{12}	-1.3649	1.1375	-2.1862***	0.7146
$\underline{q}_{ist-1} \times x_{ist}$	β_{13}	-0.4965	0.7184	-1.2393	1.0928
$\bar{q}_{ist} \times x_{ist-1}$	β_{14}	1.1417**	0.4468	0.3780	0.5049
$\underline{q}_{ist} \times x_{ist-1}$	β_{15}	0.7222	0.8475	-1.4823	0.9835
Age		-0.0365**	0.0179	-0.0247	0.0216
Years of Education		-0.0185	0.0218	-0.0190	0.0289
Married		0.1350***	0.0440	0.1056*	0.0610
Children in household		0.0255	0.0168	0.0000	0.0251
Net Income		0.0001***	0.0000	0.0001**	0.0000
Foreign		-0.1410	0.1461	-0.2112	0.2319
Private insurance		0.0543	0.0524	0.0087	0.0930
Blue collar		-0.0537	0.0387	-0.0136	0.0516
Self assessed health		0.4600***	0.0144	0.4530***	0.0160
Constant		6.7223***	0.8834	6.4740***	1.0288
α_i		yes		yes	
δ_s		yes		yes	
λ_t		yes		yes	
$(\delta_s)(\lambda_t)$		yes		yes	
N		34608		27527	

Note: *p<0.1, **p<0.05, ***p<0.01

Table 11: Employed versus unemployed – male

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/0) vs (0/1)	(1/0) vs (1/1)
(0/0)	$H_0: \beta_1 = 0$ p-value = 0.5337 $\beta_1 = -0.1930$	$H_0: \beta_1 + \beta_7 = 0$ p-value = 0.0192 $\beta_1 + \beta_7 = -1.2834$
(0/1)	$H_0: \beta_3 = \beta_1 + \beta_4$ p-value = 0.0000 $(\beta_1 + \beta_4) - (\beta_3) = -1.2070$	$H_0: \beta_3 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{15}$ p-value = 0.0023 $(\beta_1 + \beta_4 + \beta_7 + \beta_{15})$ $-(\beta_3 + \beta_{14}) = -2.7169$
(1/0)	$H_0: \beta_1 + \beta_{12} = 0$ p-value = 0.1544 $\beta_1 + \beta_{12} = -1.5579$	$H_0: \beta_1 + \beta_7 + \beta_{13} = 0$ p-value = 0.0002 $\beta_1 + \beta_7 + \beta_{13} = -1.7800$
(1/1)	$H_0: \beta_3 + \beta_8 = \beta_1 + \beta_4 + \beta_{10} + \beta_{12}$ p-value = 0.0014 $(\beta_1 + \beta_4 + \beta_{10} + \beta_{12}) - (\beta_3 + \beta_8) = -1.1335$	$H_0: \beta_3 + \beta_9 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15}$ p-value = 0.0020 $(\beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15})$ $-(\beta_3 + \beta_9 + \beta_{14}) = -0.8862$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 12: Employed versus unemployed – female

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/0) vs (0/1)	(1/0) vs (1/1)
(0/0)	$H_0: \beta_1 = 0$ p-value = 0.0767 $\beta_1 = -0.5517$	$H_0: \beta_1 + \beta_7 = 0$ p-value = 0.6326 $\beta_1 + \beta_7 = -0.4768$
(0/1)	$H_0: \beta_3 = \beta_1 + \beta_4$ p-value = 0.0009 $(\beta_1 + \beta_4) - (\beta_3) = -0.6321$	$H_0: \beta_3 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{15}$ p-value = 0.0041 $(\beta_1 + \beta_4 + \beta_7 + \beta_{15})$ $-(\beta_3 + \beta_{14}) = -2.4174$
(1/0)	$H_0: \beta_1 + \beta_{12} = 0$ p-value = 0.0000 $\beta_1 + \beta_{12} = -2.7379$	$H_0: \beta_1 + \beta_7 + \beta_{13} = 0$ p-value = 0.0003 $\beta_1 + \beta_7 + \beta_{13} = -1.7161$
(1/1)	$H_0: \beta_3 + \beta_8 = \beta_1 + \beta_4 + \beta_{10} + \beta_{12}$ p-value = 0.0870 $(\beta_1 + \beta_4 + \beta_{10} + \beta_{12}) - (\beta_3 + \beta_8) = -0.8410$	$H_0: \beta_3 + \beta_9 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15}$ p-value = 0.0003 $(\beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15})$ $-(\beta_3 + \beta_9 + \beta_{14}) = -1.9303$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 13: Expected versus unexpected unemployment – male

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/1)
(0/0) vs (1/0)	$H_0: \beta_5 + \beta_{12} = 0$ p-value = 0.2490 $(\beta_5 + \beta_{12}) = -1.3108$
(0/1) vs (1/1)	$H_0: \beta_5 + \beta_{10} + \beta_{12} = 0$ p-value = 0.7138 $(\beta_5 + \beta_{10} + \beta_{12}) = 0.1406$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 14: Expected versus unexpected unemployment – female

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/1)
(0/0) vs (1/0)	$H_0: \beta_5 + \beta_{12} = 0$ p-value = 0.0031 $(\beta_5 + \beta_{12}) = -2.1107$
(0/1) vs (1/1)	$H_0: \beta_5 + \beta_{10} + \beta_{12} = 0$ p-value = 0.7083 $(\beta_5 + \beta_{10} + \beta_{12}) = -0.1947$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 15: Fixed effects estimates for life satisfaction – by age

Variable	30 - 40		41 - 55	
	Coefficient	HAC SE	Coefficient	HAC SE
x_{ist}	β_1 -0.9223*	0.5290	-0.2328	0.2476
x_{ist-1}	β_2 -0.8241***	0.2829	0.1703	0.1578
\bar{q}_{ist}	β_3 0.2397***	0.0310	0.2118***	0.0257
\underline{q}_{ist}	β_4 -0.0409	0.5864	-0.4071	0.2838
\bar{q}_{ist-1}	β_5 0.0983***	0.0298	0.0556**	0.0259
\underline{q}_{ist-1}	β_6 0.8115***	0.2905	-0.2497	0.1747
$x_{ist} \times x_{ist-1}$	β_7 -0.8023	1.4722	-0.3582	0.5604
$\bar{q}_{ist} \times \bar{q}_{ist-1}$	β_8 -0.0961**	0.0408	0.0115	0.0361
$\bar{q}_{ist} \times \underline{q}_{ist-1}$	β_9 -2.0815***	0.7744	0.0005	0.4393
$\underline{q}_{ist} \times \bar{q}_{ist-1}$	β_{10} 2.3309***	0.8276	0.6217	1.1881
$\underline{q}_{ist} \times \underline{q}_{ist-1}$	β_{11} 0.8471	1.5446	1.0205	0.9507
$\bar{q}_{ist-1} \times x_{ist}$	β_{12} -1.9170***	0.7040	-1.1909	1.0895
$\underline{q}_{ist-1} \times x_{ist}$	β_{13} -0.0606	1.1869	-1.2872*	0.6900
$\bar{q}_{ist} \times x_{ist-1}$	β_{14} 1.9672***	0.7555	0.1033	0.3981
$\underline{q}_{ist} \times x_{ist-1}$	β_{15} 0.1054	1.6763	0.1925	0.8119
Age	-0.0467*	0.0278	-0.0147	0.0169
Years of Education	0.0003	0.0236	-0.0183	0.0280
Married	0.2077***	0.0449	0.0406	0.0659
Children in household	0.0263	0.0239	0.0277	0.0209
Net Income	0.0001***	0.0000	0.0001***	0.0000
Foreign	-0.0952	0.1671	-0.2523	0.2067
Private insurance	-0.0752	0.0653	0.1354*	0.0698
Blue collar	-0.0774*	0.0469	0.0095	0.0425
Self assessed health	0.4119***	0.0178	0.4640***	0.0136
Constant	6.7249***	1.0561	5.9936***	0.9546
α_i	yes		yes	
δ_s	yes		yes	
λ_t	yes		yes	
$(\delta_s)(\lambda_t)$	yes		yes	
N	24731		37404	

Note: *p<0.1, **p<0.05, ***p<0.01

Table 16: Employed versus unemployed – 30 - 40

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/0) vs (0/1)	(1/0) vs (1/1)
(0/0)	$H_0: \beta_1 = 0$ p-value = 0.0813 $\beta_1 = -0.9223$	$H_0: \beta_1 + \beta_7 = 0$ p-value = 0.1677 $\beta_1 + \beta_7 = -1.7246$
(0/1)	$H_0: \beta_3 = \beta_1 + \beta_4$ p-value = 0.0000 $(\beta_1 + \beta_4) - (\beta_3) = -1.2028$	$H_0: \beta_3 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{15}$ p-value = 0.0022 $(\beta_1 + \beta_4 + \beta_7 + \beta_{15})$ $-(\beta_3 + \beta_{14}) = -3.8670$
(1/0)	$H_0: \beta_1 + \beta_{12} = 0$ p-value = 0.0000 $\beta_1 + \beta_{12} = -2.8392$	$H_0: \beta_1 + \beta_7 + \beta_{13} = 0$ p-value = 0.0000 $\beta_1 + \beta_7 + \beta_{13} = -1.7852$
(1/1)	$H_0: \beta_3 + \beta_8 = \beta_1 + \beta_4 + \beta_{10} + \beta_{12}$ p-value = 0.0584 $(\beta_1 + \beta_4 + \beta_{10} + \beta_{12}) - (\beta_3 + \beta_8) = -0.6928$	$H_0: \beta_3 + \beta_9 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15}$ p-value = 0.0059 $(\beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15})$ $-(\beta_3 + \beta_9 + \beta_{14}) = -0.9989$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 17: Employed versus unemployed – 41 - 55

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/0) vs (0/1)	(1/0) vs (1/1)
(0/0)	$H_0: \beta_1 = 0$ p-value = 0.3472 $\beta_1 = -0.2328$	$H_0: \beta_1 + \beta_7 = 0$ p-value = 0.2494 $\beta_1 + \beta_7 = -0.5910$
(0/1)	$H_0: \beta_3 = \beta_1 + \beta_4$ p-value = 0.0000 $(\beta_1 + \beta_4) - (\beta_3) = -0.8517$	$H_0: \beta_3 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{15}$ p-value = 0.2085 $(\beta_1 + \beta_4 + \beta_7 + \beta_{15})$ $-(\beta_3 + \beta_{14}) = -1.1207$
(1/0)	$H_0: \beta_1 + \beta_{12} = 0$ p-value = 0.1787 $\beta_1 + \beta_{12} = -1.4238$	$H_0: \beta_1 + \beta_7 + \beta_{13} = 0$ p-value = 0.0000 $\beta_1 + \beta_7 + \beta_{13} = -1.8782$
(1/1)	$H_0: \beta_3 + \beta_8 = \beta_1 + \beta_4 + \beta_{10} + \beta_{12}$ p-value = 0.0019 $(\beta_1 + \beta_4 + \beta_{10} + \beta_{12}) - (\beta_3 + \beta_8) = -1.4324$	$H_0: \beta_3 + \beta_9 + \beta_{14}$ $= \beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15}$ p-value = 0.0000 $(\beta_1 + \beta_4 + \beta_7 + \beta_{11} + \beta_{13} + \beta_{15})$ $-(\beta_3 + \beta_9 + \beta_{14}) = -1.3879$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 18: Expected versus unexpected unemployment – 30 - 40

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/1)
(0/0) vs (1/0)	$H_0: \beta_5 + \beta_{12} = 0$ p-value = 0.0098 $(\beta_5 + \beta_{12}) = -1.8187$
(0/1) vs (1/1)	$H_0: \beta_5 + \beta_{10} + \beta_{12} = 0$ p-value = 0.2477 $(\beta_5 + \beta_{10} + \beta_{12}) = 0.5123$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 19: Expected versus unexpected unemployment – 41 - 55

$(q_{t-1}/q_t) \backslash (x_{t-1}/x_t)$	(0/1)
(0/0) vs (1/0)	$H_0: \beta_5 + \beta_{12} = 0$ p-value = 0.2972 $(\beta_5 + \beta_{12}) = -1.1354$
(0/1) vs (1/1)	$H_0: \beta_5 + \beta_{10} + \beta_{12} = 0$ p-value = 0.2820 $(\beta_5 + \beta_{10} + \beta_{12}) = -0.5137$

Note: $q_t = 1$ if expectations are positive, $x_t = 1$ if unemployed in t

Table 20: Interpretation of coefficients

	q_{it-1}	0		1	
x_{it-1}	q_{it} x_{it}	0	1	0	1
0	0	⁽⁰⁰⁰⁰⁾ β_0	⁽⁰⁰⁰¹⁾ $\beta_0 + \beta_3$	⁽⁰⁰¹⁰⁾ $\beta_0 + \beta_5$	⁽⁰⁰¹¹⁾ $\beta_0 + \beta_3 + \beta_5 + \beta_8$
	1	⁽⁰¹⁰⁰⁾ $\beta_0 + \beta_1$	⁽⁰¹⁰¹⁾ $\beta_0 + \beta_1 + \beta_4$	⁽⁰¹¹⁰⁾ $\beta_0 + \beta_1 + \beta_5 + \beta_{12}$	⁽⁰¹¹¹⁾ $\beta_0 + \beta_1 + \beta_4 + \beta_5$ $+ \beta_{10} + \beta_{12}$
1	0	⁽¹⁰⁰⁰⁾ $\beta_0 + \beta_2$	⁽¹⁰⁰¹⁾ $\beta_0 + \beta_2 + \beta_3 + \beta_{14}$	⁽¹⁰¹⁰⁾ $\beta_0 + \beta_2 + \beta_6$	⁽¹⁰¹¹⁾ $\beta_0 + \beta_2 + \beta_3 + \beta_6 + \beta_9$ $+ \beta_{14}$
	1	⁽¹¹⁰⁰⁾ $\beta_0 + \beta_1 + \beta_2 + \beta_7$	⁽¹¹⁰¹⁾ $\beta_0 + \beta_1 + \beta_2 + \beta_4 + \beta_7$ $+ \beta_{15}$	⁽¹¹¹⁰⁾ $\beta_0 + \beta_1 + \beta_2 + \beta_6 + \beta_7$ $+ \beta_{13}$	⁽¹¹¹¹⁾ $\beta_0 + \beta_1 + \beta_2 + \beta_4 + \beta_6 + \beta_7$ $+ \beta_{11} + \beta_{13} + \beta_{15}$

Note: $q_{it} = 1$ if expectations are positive, $x_{it} = 1$ if unemployed in t

Table 21: Summary statistics for life satisfaction and expectations by employment status and age

Age	Variable	$x_{it} = 0$			$x_{it} = 1$		
		Mean	SD	N	Mean	SD	N
30	y	7.255	1.523	1662	5.824	2.430	17
	\bar{q} / \underline{q}	0.425	0.495		0.882	0.332	
31	y	7.276	1.488	1729	5.222	2.108	9
	\bar{q} / \underline{q}	0.426	0.495		0.778	0.441	
32	y	7.229	1.536	1871	6.524	2.205	21
	\bar{q} / \underline{q}	0.433	0.496		0.905	0.301	
33	y	7.198	1.527	2022	6.111	1.779	18
	\bar{q} / \underline{q}	0.409	0.492		0.944	0.236	
34	y	7.209	1.500	2140	5.050	1.820	20
	\bar{q} / \underline{q}	0.422	0.494		0.800	0.410	
35	y	7.192	1.515	2275	5.583	1.311	11
	\bar{q} / \underline{q}	0.411	0.492		0.667	0.492	
36	y	7.157	1.526	2404	5.294	2.085	17
	\bar{q} / \underline{q}	0.395	0.489		0.941	0.243	
37	y	7.124	1.558	2464	5.867	1.767	15
	\bar{q} / \underline{q}	0.395	0.489		0.933	0.258	
38	y	7.060	1.574	2583	5.889	1.530	17
	\bar{q} / \underline{q}	0.391	0.488		0.611	0.502	
39	y	7.124	1.560	2672	5.125	1.893	16
	\bar{q} / \underline{q}	0.397	0.489		0.813	0.403	
40	y	7.095	1.563	2728	5.611	3.183	18
	\bar{q} / \underline{q}	0.391	0.488		0.833	0.383	
41	y	7.027	1.565	2785	5.500	1.713	16
	\bar{q} / \underline{q}	0.386	0.487		0.875	0.342	
42	y	7.048	1.563	2836	5.692	2.175	13
	\bar{q} / \underline{q}	0.384	0.486		0.769	0.439	
43	y	6.960	1.620	2787	5.000	1.440	25
	\bar{q} / \underline{q}	0.375	0.484		0.750	0.441	
44	y	6.970	1.629	2792	5.462	1.964	24
	\bar{q} / \underline{q}	0.371	0.483		0.731	0.452	
45	y	6.961	1.609	2727	4.762	1.921	21
	\bar{q} / \underline{q}	0.371	0.483		0.762	0.436	
46	y	6.971	1.617	2630	5.294	1.993	16
	\bar{q} / \underline{q}	0.390	0.488		0.765	0.437	
47	y	6.965	1.662	2595	5.360	2.378	23
	\bar{q} / \underline{q}	0.378	0.485		0.600	0.500	
48	y	6.953	1.681	2568	5.261	2.220	23
	\bar{q} / \underline{q}	0.385	0.487		0.783	0.422	
49	y	7.010	1.624	2487	5.053	2.041	19
	\bar{q} / \underline{q}	0.403	0.491		0.895	0.315	
50	y	6.995	1.637	2416	5.217	1.882	21
	\bar{q} / \underline{q}	0.404	0.491		0.609	0.499	
51	y	6.998	1.655	2324	4.895	1.792	18
	\bar{q} / \underline{q}	0.402	0.490		0.684	0.478	
52	y	6.994	1.623	2227	5.526	1.806	19
	\bar{q} / \underline{q}	0.431	0.495		0.684	0.478	
53	y	7.033	1.625	2127	5.500	1.900	10
	\bar{q} / \underline{q}	0.455	0.498		0.500	0.527	
54	y	7.016	1.620	1989	6.400	1.776	10
	\bar{q} / \underline{q}	0.454	0.498		0.700	0.483	
55	y	6.972	1.646	1838	6.143	1.864	6
	\bar{q} / \underline{q}	0.477	0.500		0.429	0.535	

y : life satisfaction (0 = low, 10 = high)

\bar{q} : share of employed with positive expectations

\underline{q} : share of unemployed with positive expectations

Table 22: Counts of correctly predicted unemployment - by age

	30 - 40		41 - 55	
	$x_{it} = 0$	$x_{it} = 1$	$x_{it} = 0$	$x_{it} = 1$
$\bar{q}_{it-1} = 1$	10 086	22	14 886	10
$\bar{q}_{it-1} = 0$	13 796	93	21 490	180
Total	23 882	115	36 376	190
$\underline{q}_{it-1} = 1$	585	50	585	60
$\underline{q}_{it-1} = 0$	83	16	167	26
Total	668	66	752	86

\bar{q} : expectations of the employed

\underline{q} : expectations of the unemployed

Table 23: Counts of unemployment periods - by age groups

Counts $\sum_{t=1}^T x_{it}$	30 - 40		41 - 55	
	Number	Percent	Number	Percent
0	24 081	97.37	36 372	97.24
1	517	2.09	802	2.14
2	107	0.43	175	0.47
3	20	0.08	48	0.13
4	6	0.02	7	0.02
Total	24 731	100.00	37 404	100.00