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by

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The Persistence of Welfare Participation

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September 2007

Abstract:

Welfare persistence is estimated in and compared between Swedish-born and foreign-born households. This is done within the framework of a time-stationary dynamic discrete choice model controlling for the initial condition and unobserved heterogeneity. Three different types of persistence are controlled for in terms of observed and unobserved heterogeneity, serial correlation, and structural state dependence, the focus being on the latter measure. In a second step we analyze the long-run effects of receiving social assistance on future household earnings and disposable income.

The results show that state dependence in Swedish welfare participation is strong in both Swedish-born and foreign-born. However, the size of the effect is three times as large for the latter group. When the effect is distributed over time, it disappears after three years for both groups. The effect of structural state dependence is decomposed into a number of observed explanatory factors. Surprisingly small effects are found from typical foreign-born factors such as time in the country and country of origin, both important determinants for welfare participation in general.

When investigating the effect of social assistance participation on future earnings, we find a strong and persistent effect over the whole observation window, while no such effect could be found for disposable income. This indicates that the economic incentives to leave the dependency are very weak. The picture is similar for both Swedish-born and foreign-born, even though the negative earnings effect is somewhat larger for the latter.

Kewords: welfare participation, immigrants, dynamic probit model, persistence, state dependence, unobserved heterogeneity, initial condition, GHK simulator, earnings, disposable income

JEL Classification: I30, I38, J18

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1. Introduction

The expenditure on welfare increased substantially in many western countries during the 1990s, and the Scandinavian countries are in the top in terms of public spending on social support. Even though pensions and health care stand for the largest part of public social spending, income support to the working-age population accounts for a major and currently increasing part (Adema, 2006). Expenditures related to welfare participation in terms of social assistance, is therefore a problem attracting special attention by many European governments.

In Sweden, the official statistics show that the total stock of social assistance recipients decreased during the end of the 1990s and the first half of the 2000s, which is believed to be a result of the improved labor market conditions. However, the long-term participants were not significantly affected by the positive labor market changes. On the contrary, their share is currently increasing and now stands for a major part of the expenditure; consequently, long-term welfare participation is an important economic issue for policy makers to deal with.

In order to reduce this problem it is important to understand both the mechanisms that drive people into welfare, and the causes that make some people stay on welfare for long periods of time. In the literature it is often noted that individuals with previous experience of welfare have an increased risk of future participation. An explanation for this observed event might be that it is the experience in itself that alters the cost or the stigma related to welfare participation, shifting the structure of the individual's preferences and in the end increasing the likelihood of remaining on welfare in the following period. If this is true, efforts should be made to avoid short-term economic policies that increase people's likelihood of being exposed to this experience.

An alternative explanation could be that the observed persistence is due to innate individual differences, and that some individuals have a larger propensity to live on welfare than others. If these differences among people are not properly controlled for when describing the patterns of welfare participation, then the observed persistence will be due to a conditional dependency between past and present, which doesn't need to be related to preferential changes in the individual.

The international literature on welfare participation is vast (for a summary, see Danziger et al., 1981; Lichter et al., 1997; and Moffit, 1992, 1998). However, the body of literature focusing on state dependence and social assistance is very small. Hansen and Lofstrom (2003, 2006) are two studies with a setup similar to ours

focusing on differences in welfare participation between immigrants and natives. They found differences in welfare participation between the groups and suggest that an important factor behind this may be entry rates. Another study by Hansen et al. (2006) analyzes the transition into and out of social assistance in Canada, finding that there are substantial differences across different provinces.

We extend this literature and focus on how the shape and the persistence of state dependence affect the persistence in welfare participation. The structure of this behavior is important to understand in order to be able to increase the outflow and reduce the inflow of newcomers. The aim of this study is therefore to analyze the importance and the size of the effect of structural state dependence in welfare participation, and investigate how observed factors (individual and macro-related factors) are associated with this dependence and how it persists over time.

In order to study the "true" state dependence it is necessary to investigate the dynamic structure of participation, accounting for unobserved individual differences and separating them from a possible state dependence. This will be done using a general time stationary dynamic discrete choice model proposed by Heckman (1981a). It incorporates state dependence while controlling for the initial condition problem and for individual unobserved heterogeneity using a general intertemporal covariance structure. The analysis is done separately for Swedish-born and foreign-born people during 1990-1999 period.

The rest of the paper is organized in the following way. The next section describes the welfare system in Sweden in the analyzed period. The empirical specification and the estimation method are described and discussed in Section 3. The data is presented in Section 4, and Section 5 discusses and analyzes the results from the empirical model. Section 6 sums up and concludes the paper.

2. Welfare participation and persistence in Sweden

Social assistance is the final safety net for households that have run out of financial means to maintain their daily livelihood. The Swedish law gives all households the right to a minimum standard of living, implying that if a person is completely without any financial means, the state will pay for an apartment, childcare, food etc. with the requirement that the welfare applicant makes a full-time effort to find a job, or to receive income from other sources. This means that the applicant cannot voluntarily

give up a job in order to live on social assistance. Furthermore, personal assets (with some exceptions) must be spent before any social assistance may be received.

The total welfare benefit offered by the state may be decomposed into two parts. The first part is a regulated component that pays for housing, childcare, and similar expenses. The second part is meant to cover the more basic daily consumption needs of the household, such as food and clothing. The level of the second component is referred to as the social assistance norm and is regulated by the welfare recipient's home municipality. The National Board of Health and Welfare provide guidelines to the municipalities in order to harmonize the level across the country.

The assistance application process takes place at the social welfare office, typically on monthly basis. It is the individual who chooses to visit the welfare office, while a social worker decides whether the household is entitled to welfare benefits. The decision is based on an interview process, going through the financial situation of the household.

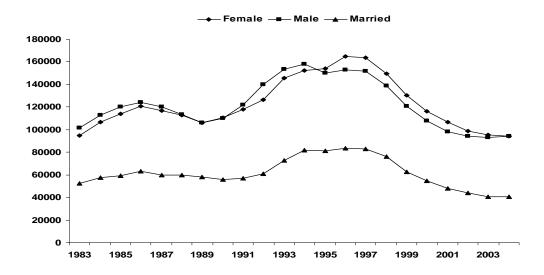


Figure 1 Number of welfare participants by household type, 1993 - 2004 Source: Statistics Sweden

As can be seen in Figure 1, the social assistance rate changed dramatically during the 1983-2004 period, with a peak in 1997. The participation rate increased by almost 50% from 1990 to 1997, but then decreased steadily until 2004.

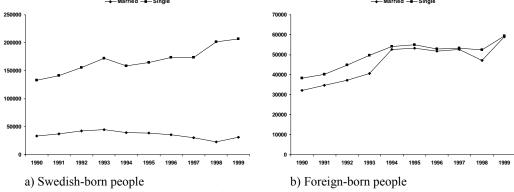


Figure 2 Welfare participation over time by household type in level Source: The Swedish Income Panel (SWIP), 1990 – 1999

The participation rate differs substantially among different groups of people which make it interesting to disaggregate the information given in Figure 1 to receive a better picture of the participation behavior. Figure 2 splits the welfare recipients into Swedish-born and foreign-born and into single and non-single households, making differences appear: Among the Swedish-born a clear majority of the welfare recipients are single-person households. A majority of those single-person households are single men with no children and single women with children. Foreign-born households are different in this respect, showing much smaller differences: from 1994 to 1997 there were basically no differences between the household types. From 1994 and on, the foreign-born welfare households with several members out-number the Swedish-born counterparts in absolute terms, which clearly states the difference since only 9-11% (1990-2000) of the Swedish population are foreign-born.

When looking at the relative change in participation rates, differences between Swedish and foreign-born people once again appear. The major recession in the Swedish economy started in 1990-1991, and the changes in participation were basically the same until 1993, as can be seen in Figure 3.

The labor market conditions improved slightly in 1994, which affected the change in participation rate for the Swedish-born but not for the foreign-born. This is an indication of a lag in response to labor market conditions compared to the Swedishborn. This could also be an indication that the state dependence is much stronger for foreign-born people.

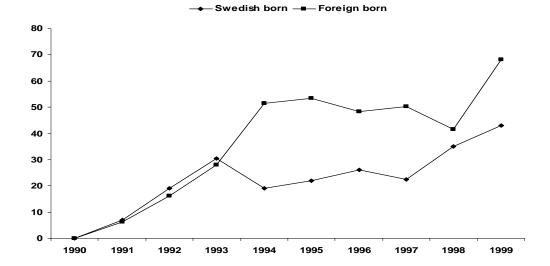


Figure 3 Percentage change in welfare participation 1990-1999 Source: The Swedish Income Panel (SWIP), 1990 – 1999

Figure 4 presents relative change in participation rate for different household types. Large differences can be seen. For the Swedish-born, the change in participation rate is initially about the same for single-person households and households with several persons. After 1994 something happens, and there is a major drop for cohabitants, while the change in participation rate for single person households continues to increase.

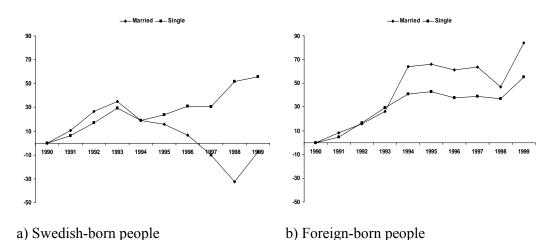
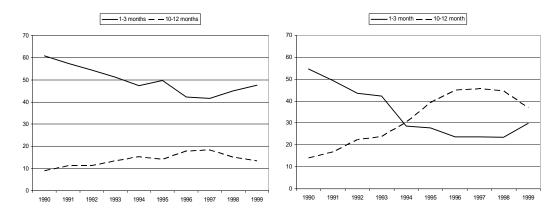


Figure 4 Percentage change in welfare participation 1990-1999 by household type Source: The Swedish Income Panel (SWIP), 1990 – 1999.

For the foreign-born group, the growth in the participation rate is about constant until 1993. From 1994 and on, the cohabitant level of change increases at a much steeper

pace. The data used in Figure 4 corresponds to the data used in Figure 2, and it can be seen that the increased growth rate among the cohabitants resulted in the two groups (married and single) become just about equal in level.

The participation rate has so far been described on a yearly basis, where a person is defined as a welfare participant if he or she has received social assistance in at least one month during a given year. However, since the actual participation decision is made on a monthly basis, further insight into welfare participation behavior could be gained by analyzing the number of months people receive social assistance during a year.



- a) Swedish-born welfare participants
- b) Foreign-born welfare participants

Figure 5 Short-term and long-term welfare participation between 1990-1999 (in percent)

Official statistics show that most welfare participants are short-time social assistance receivers, receiving support only one or two months in a year. However, during the 1990s the proportion changed, especially for the foreign-born. Figure 5 describes the general picture for Swedish-born and foreign-born welfare participants, and how the shares of short-term and long-term participants changed over the decade. By convention, a long-term participant is defined as having received social assistance for ten months or more in a year, and we can see that this proportion increased extensively during the 1990s. This is especially true for the foreign-born. After 1990, when the shares were about the same for the two groups, they both started to grow. However, the long-term share of the foreign-born grew much faster, eventually surpassing the short-term receivers in 1994. This dramatic increase for the foreign-born can to a large extent be explained by the economic recession and the large inflow

of refugees in the mid 1990s. Apparently there are great differences between the two groups in terms of participation behavior. The interesting question is whether this is due to behavioural differences on an individual basis or if it is related to structural factors?

3. The Empirical Specification

The point of departure is an economic agent with perfect foresight that in each time period makes a decision about welfare participation with the objective of maximizing his expected lifetime utility. Each decision is discrete, so that within each time period there is no decision about being on welfare. Even though the decision is discrete, it is based on a latent continuous measure Y_{it} *, representing the individual propensity to participate. This measure is a construction of the difference between the individual utility with and without welfare in a given time period. Whenever the utility with welfare is greater then the utility without welfare, an individual will choose the welfare alternative. Hence, it is the difference in utilities that is the relevant measure when an individual is making a decision. However, an individual's current utility difference is a function of the difference in the previous period. The difference in period t may therefore be decomposed in the following way:

$$Y_{it}^* = X_{it}\beta + \sum_{j=1}^{s} \gamma_j Y_{it-j} + v_{it} \text{ with } Y_{it} = \begin{cases} 1 & Y_{it}^* \ge 0 \\ 0 & Y_{it}^* < 0 \end{cases}.$$

$$i = 1, ..., N, \text{ and } t = 1, ..., T$$

$$(1)$$

The error term v_{it} is assumed to be independent of X_{it} and is independently distributed over i. Within the observations of each individual, v_{it} is assumed to be distributed multivariate normal with a mean zero and a general intertemporal covariance matrix Ω . The availability of panel data provides for the possibility to distinguish average behavior from individual behavior by decomposing the effects of the error term v_{it} into $v_{it} = f(\alpha_{i}, u_{it})$, where α_{i} denotes the effect of omitted individual specific variables and u_{it} is a residual term representing effects of factors other than the individual specific characteristics not observed by the investigator. Hence, the existence of an individual specific unobserved permanent component allows individuals who are homogenous in their observed characteristics to be heterogeneous in their response variables. This model is consistent with McFadden's (1973) random utility model applied in an intertemporal context, given the assumptions made here.

3.1 Welfare persistence

Specification (1) allows for three different sources of persistence after controlling for observed explanatory factors. Persistence can be a result of serial correlation in the error term, u_{it} , a results of unobserved heterogeneity, α_i , or a result of "true" or "structural state dependence" through the term $\sum_{j=1}^{s} \gamma_j Y_{i,t-j}$. Although all three sources are interesting, the focus will be on the size and distribution of the components of the "true" state dependence, while controlling for the other two sources. If the components in the intertemporal covariance matrix are significantly different from zero, then unobserved individual specific heterogeneity and serial correlation will affect the estimates for the state dependence if not controlled for.

As indicated, the existence of a "true" state dependence will be tested in this study. The measure γ captures the idea that the effect of an experience in the previous period has a real and behavioral effect on the choice in the current period. In a first step the structure is limited to a first order Markov process that captures the correlation between pairwise observations over time. $\gamma > 0$ would imply that the likelihood of being dependent on welfare in the current period is larger for those with an earlier experience compared to others without such an experience. In a second step we relax the assumption of a first order Markov process and allow for more lags; we can then see how many years it takes to lose the increased risk of returning to welfare dependency as a result of the first initial experience.

To investigate the factors affecting the first order state dependence, the overall effect will be decomposed into several observed explanatory factors that potentially affect the size of the state dependence. That is, a linear approximation will be applied in the following way: $\gamma = z\delta$, with z being a vector of observed factors, and δ being a vector of parameters. With this specification, a deeper understanding of the factors behind the event can be gained.

Distinguishing between structural (true) and spurious state dependence is of considerable interest, since they have very different policy implications. A policy that temporarily increases the probability of participation has different implications for future probabilities in a model with true state dependence than in a model where the persistence is due to unobserved heterogeneity and/or serial correlation.

3.2 Estimation and identification

The estimation method applied in this study is based on the maximum likelihood technique, which requires the formulation of a likelihood function. The model as described by equation (1) is based on ten time periods and results in a log-likelihood function in the following way:

$$L = \sum_{i=1}^{N} \log[prob(Y_{i1}, Y_{i2}, ..., Y_{i10})], \qquad (2)$$

where

$$Prob(Y_{i1}, Y_{i2}, ..., Y_{i10}) = \int_{a_{i1}}^{b_{i1}} ... \int f(v_{i1}, ..., v_{i10}) dv_{i10}, ..., dv_{i1}$$

 $a_{it} = -X_{it}\beta$ and $b_{it} = \infty$ if $Y_{it} = 1$, while $a_{it} = -\infty$ and $b_{it} = -X_{it}\beta$ if $Y_{it} = 0$. f(.) is the multivariate normal density function. The standard difficulty in this problem is the evaluation of the ten fold integral in equation (2), which will be solved using a smooth recursive conditioning simulator that simulates the multivariate probabilities rather than evaluating them numerically. The GHK recursive simulator, (Geweke, 1991; Hajivasssiliou and McFadden, 1990; and Keane, 1990, 1994) is based on the observation that the choice probabilities in the multinomial probit model can be written as a sequence of conditional probabilities that may be simulated recursively. This simulator is of particular interest because it has been shown in a rather exhaustive study of many alternative probability simulators by Hajivassiliou, McFadden, and Ruud (1996) to be the most accurate and reliable simulator of all those considered (see also Gouriéroux and Monfort, 1993; and Keane, 1993, which focus explicitly on applications of simulation methods to panel data). An additional beneficial feature of the GHK simulator is that it is rather easy to implement for this kind of model. The likelihood function described above may therefore be rewritten as:

$$L_{SML} = \frac{1}{R} \sum_{i=1}^{R} \prod_{t=1}^{T} Q_{t} (\eta_{1}^{r}, ..., \eta_{t-1}^{r}),$$
(3)

where $\prod_{t=1}^{T} Q_t$ represents the sequence of conditional probabilities, and η_t^r the random draws from the truncated normal density (for an intuitive description of the procedure, see Train, 2003). The simulated likelihood is a continuous and differentiable function of the parameters to be estimated. In addition, the simulated

likelihood function is an unbiased estimator of the likelihood function (Börsch-Supan and Hajivassiliou, 1993). However, in order to receive consistency in the simulated estimation, the number of simulated draws R has to be large enough. Under certain regularity conditions, a sufficient rate is $R/\sqrt{N} \to \infty$ as $N \to \infty$ in order to obtain consistent, asymptotically normal and efficient estimates (Hajivassiliou and Ruud, 1994).

Since this is a dynamic model, two additional complications need to be solved in order to receive consistent estimates of the parameters of interest: the initial condition problem and the necessity of separating the effect of unobserved individual characteristics from the possible effect of state dependence. The first problem is related to the fact that we are unable to observe the data generating process from its beginning. In the sample of individuals used there are some with previous experience of welfare participation who are not accounted for in the initial year of the observed series, which generates a conditional relationship causing inconsistent estimates of the parameters of interest. If the process is in equilibrium or if the previous experience is independent and exogenous of the behavior observed during the first time period, then there is no problem. However, assuming this to be the case would be unreasonable. The problem of the initial condition declines with the length of the panel, but the panel length in this study is only ten time periods, something that requires special attention. Heckman (1981b) proposes a statistical approximation method that solves the problem with reasonable precision.² This is done by approximating the initial state in the sample using a univariate probit model, estimating its parameter separately and allowing its error term to freely correlate with the error terms of the remaining time periods and thereby circumvent the endogeneity problem.

The second problem to consider is the problem of distinguishing between true and spurious state dependence, which is the same as separating the effects of unobserved individual characteristics from the potential effect of state dependence. This problem and its solutions are related to the assumptions made on the residual term in equation (1). In the literature there are many examples of more or less restrictive ways of dealing with the residual term in order to separate out the individual specific effects. Two alternative specifications will be used here to single out which structure best describes the data and how the structures compare to each other. The most general

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² See Wooldridge (2005) for an alternative method.

form is based on a free covariance structure and represents a very general structure for the residual term. The more restricted structures that are to be compared with this general model are (1) a first-order Markov process $(v_{ii} = \rho v_{i,t-1} + u_{ii})$ allowing for serial correlation and assuming that no other effects remain in the residual term, and (2) a conventional component of variance scheme $(v_{ii} = \alpha_i + u_{ii})$, which is very often used in the literature.³

In order to be able to estimate and identify the parameters of the main model, it is important to impose a couple of normalizations. For the coefficients of the model to be consistently estimated, it is sufficient to normalize the variance of the first time period only, which means that it is possible to allow for heteroscedasticity over time. However, when using the GHK simulator, such normalization causes an asymmetry in the simulated error structure, biasing the standard errors (for the coefficients of the remaining time periods) received from the estimated information matrix using standard numerical methods such as the finite difference approach. Therefore, the variances for all time periods have been normalized to one, imposing homoscedasticity over time. However, when testing for this restriction, it turned out not to be a problem, since any deviation from homoscedasticity was absorbed by the remaining free components of the covariance matrix. The information matrix is approximated using the BHHH method.

4. Data

We have access to a register database (SWIP) that constitutes a stratified random sample of the population living in Sweden.⁴ It is stratified into two parts: the first is a 1% sample of the Swedish-born population and the second is a 10% sample of the foreign-born population living in Sweden. The stratified random sample was drawn by Statistics Sweden using population files from 1978 and on. The individuals in the initial year were followed over time with repeated yearly cross-sections. To each consecutive year a supplement of individuals was added to each cross-sectional unit in order to adjust for migration and those born since the last survey (previous year); the

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 $^{^3}$ Specification (1) corresponds to $\rho_{t,t+k}=\rho^k$, and (2) corresponds to $\rho_{t,t+k}=\sigma_\alpha^2/1+\sigma_\alpha^2$.

⁴ The Swedish Income Panel (SWIP) is a register-based panel data set administrated by the Swedish Social Science Data Service (SSD). More information can be found at www.ssd.gu.se.

intention being to make each stratified cross-section representative of the Swedish population with respect to each stratum.

This construction makes it possible to follow individuals over time and the analysis is based on a random sample of the working population in 1990, aged 18-50, excluding students and retired people. Social assistance is applied for by the household, and it is consequently the household as a unit that is analyzed by the social worker when investigating eligibility. In the literature it is often the ambition to describe the household by the characteristics of the household head. Unfortunately, SWIP offers no such information, and hence we are unable to identify the household head. As a substitute for the household head we use the characteristics of the sample person. That is, age, education and so forth are factors related to the person originally sampled into SWIP. Furthermore, the stratified random samples as given would have resulted in very large data sets when considering the full time period, which led us to reduce the sample to around 10,000 individuals in the initial year of 1990. In order to balance the panel, some individuals had to be dropped; the final samples of individuals were reduced to 8,205 and 8,407 for the Swedish-born and the foreign-born, respectively.

4.1 Variable definitions and characteristics

A household is defined as a social assistance recipient if the sample person of that household has received social assistance at least once during a calendar year. As described in the data section, this aggregated design implies that some information on welfare behavior is lost. There is unfortunately nothing we can do to change that, since we have no information about the sequence of social assistance, received during the year, just the number of months.

The variables used as observed explanatory factors in the analysis are presented in Table 1, which shows mean values for the whole period. Comparing Swedish- and foreign-born we see that average age is about the same, while a relatively larger proportion of the Swedish-born social assistance recipients are found in the youngest age category.

The educational level of the two groups does not differ much, even though there is a slight concentration on secondary schooling for the Swedish-born and primary schooling for the foreign-born. The number of children is usually a factor that is related to welfare recipients, especially when the children are younger. We can see that the shares of Swedish- and foreign-born households with children younger than

six are about equal in size. However, if we look at foreign-born welfare recipients, then this is much larger than among the Swedish-born. Hence, the presence of young children in the household seems to be a more important factor among the foreign-born.

Table 1 Mean observable characteristics of welfare recipients, 1990 – 1999

	Sı	vedish-born po	eople	Foreign-born people			
	Total	Welfare	No welfare	Total	Welfare	No welfare	
		recipients	recipients		recipients	recipients	
Age	39.9	35.3	40.0	40.1	36.6	40.5	
Age 18-30 (%)	19.1	35.0	18.6	15.6	24.7	14.5	
Age 31-40 (%)	31.5	35.1	31.4	35.2	42.9	34.2	
Age 41-50 (%)	49.4	29.9	50.0	49.2	32.4	51.2	
Educational level							
Primary school (%)	22.8	44.1	22.1	37.2	53.9	35.2	
Secondary School (%)	51.2	51.8	51.2	42.6	36.6	43.3	
Post secondary School (%)	25.9	4.1	26.7	20.2	9.4	21.5	
Children aged less than 6 (%)	16.7	17.2	16.7	17.4	24.7	16.5	
Cohabitant (%)	59.8	21.2	61.2	61.4	43.0	63.6	
City region (%)	24.5	30.2	24.3	36.6	45.4	35.6	
Unemployed (%)	13.5	40.9	12.5	20.9	37.4	18.9	
Regional rate of welfare participation	4.9	5.4	4.8	5.6	5.9	5.5	
SA norm	7797	7977	7791	8008	8011	8008	
Average regional welfare duration	4.6	4.6	4.6	6.6	6.7	6.6	
Sample size	82050	2738	79312	84070	9138	74932	

Source: The Swedish income panel (SWIP)

Cohabitation is another factor that is related to welfare dependence. It is often noted in the literature that the event of a divorce or a separation is an important route into welfare dependency. However, while this might be the case for the Swedish-born group, the situation is somewhat different for the foreign-born. This is due to the proportion of married welfare recipients being twice as large among the foreign-born and that during the mid 1990s the number of married welfare recipients increased substantially. The negative married effect on welfare participation should therefore be much smaller in the foreign-born group.

Large city region is another factor that might be related to welfare participation. We know that in general, foreign-born people choose to live in larger cities (this is confirmed by the descriptive statistics above). Hence, foreign-born welfare recipients

also tend to live in large city regions to a large extent: 45% compared to 30% among the Swedish-born.

Unemployment is another important factor that explains whether households end up on welfare, and we observe that around 40% of the welfare recipients in both groups have been unemployed during the year. The unemployment variable is binary, and an individual is defined to have been unemployed during a year if he has received any cash assistance or unemployment insurance during the year. This implies that a sample person could have been unemployed only very briefly during the year. The construction also implies that we miss those who are unemployed and are not entitled to cash assistance or unemployment insurance, which is a group that to a large extent are directed to social assistance. People not entitled to unemployment benefits are usually very young and without previous work experience, since eligibility for unemployment benefit usually requires some work history.

Regional rate of welfare participation is a variable that is based and constructed on sub-groups of welfare recipients who appear in SWIP using the full sample of the year. Hence, for the Swedish-born we calculate the average participation rate for each municipality in Sweden. This variable is based on the idea that households in a municipality with a large number of welfare recipients, and that are at the margin of being a welfare recipients themselves are more likely to take the step into the welfare office compared to households in other areas. On the other hand, there might be alternative explanations. It is quite likely that welfare office generosity differs among municipalities and that it is easier to receive social assistance in some places, which therefore generates a positive relationship between welfare participation and the average regional participation rate. At any rate, we are unable to differentiate between these two effects in the model.

Another structural variable is the social assistance norm. Unfortunately, we do not have access to the norm for each municipality, so we have to create a proxy. It is reasonable to believe that the norm is related to the disposable income of welfare recipients. We therefore calculate the average disposable income of welfare recipients in each of the municipalities in Sweden and over time, using the full sample of SWIP. Hence, if the disposable income of welfare recipients increases, it is plausible to believe that the social assistance norm have increased as well, which means that this proxy should work well. This implies that if the norm is increasing, more people

should be eligible for social assistance, and hence more people will receive social assistance.

The average regional welfare duration is related to the regional rate of welfare participation, but the link is not obvious. There could be regions with low rates of participation, but with longer welfare spells. There could also be other combinations. When looking at a simple unconditional correlation measure, we find a positive relation although the correlation is weak, which implies that the dispersion is great. It is therefore difficult to say if and how average regional welfare duration is related to the welfare participation rate.

Table 2 Welfare participation by cohorts over time: 1990 – 1999

	Pa	rticipation rate (%)	
	1990	1995	1999	Percentage difference 1990-1999
Swedish-born cohorts				
All	3.4	3.5	2.4	-29.4
Age $(18 - 30)$	5.5	4.9	3.4	-38.2
Age $(31 - 40)$	2.9	4.0	2.7	-6.9
Age (41 – 50)	1.9	1.6	1.1	-42.1
Foreign-born cohorts				
All	12.7	10.4	8.3	-34.6
Age $(18 - 30)$	18.6	14.0	10.9	-41.4
Age $(31 - 40)$	13.3	11.6	9.0	-32.3
Age $(41 - 50)$	6.3	5.6	4.8	-23.8
Years in the country, (in 199	90)			
0 - 4	29.9	20.5	15.3	-48.8
5 – 9	12.0	11.8	10.0	-16.7
10 - 14	7.4	8.1	6.9	-6.8
15 - 22	6.5	6.4	5.6	-13.8
>22	3.9	4.2	3.2	-17.9
Country of origin ⁵				
Nordic country	7.1	6.8	5.3	-25.4
Western Europe	3.5	3.9	2.5	-28.6
Eastern Europe	12.7	7.4	6.1	-52.0
Southern Europe	5.4	6.1	5.3	-1.9
Middle East	30.2	22.6	19.9	-34.1
Rest of the world	22.4	18.0	12.3	-45.1
Refugee country ⁶	20.8	15.5	12.5	-39.9

Source: The Swedish income panel (SWIP).

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⁵ Categories: Nordic (Denmark, Finland, Norway, Iceland), Western Europe (Germany, France, Benelux, Switzerland, Austria, UK, Ireland), Eastern Europe (Poland, Hungary, Albania, Bulgaria, Romania, Czechoslovakia, countries in the former Soviet Union), Southern Europe (Greece, Yugoslavia, Andorra, Italy, Portugal, San Marino, Spain, the Vatican state), Middle East (Arab countries, Iraq, Iran, Turkey), and the rest of the world.

⁶ Refugee countries according to the Swedish Immigration Board: Afghanistan, Bangladesh, Bosnia, Bulgaria, Chile, Cuba, China, Croatia, Ethiopia, India, Iran, Iraq, Sri Lanka, Lebanon, Moldavia, Pakistan, Peru, Poland, Romania, Russia, other states of the former Soviet Union, Somalia, Syria, Togo, Turkey, Ukraine, Uganda, Vietnam, and Yugoslavia.

For the foreign-born we have a group of variables that are important for success on the labor market, namely the number of years in the country, the country of origin, and whether or not the country of origin is a refugee country; that is, whether or not the person arrived in Sweden as a refugee.

Table 2 shows the participation rates for different sub-groups of the 1990 cohort, and then how they change over time. The first part is related to the Swedish-born group, where participation rate has been calculated for three age groups. In the initial year, we see that the youngest group had the largest participation rate. We know that the participation rate increased from 1990 until 1997 for the group as a whole. However, there are differences between the age groups: For the youngest group, the participation rate decreased between 1990 and 1995, while it at the same time increased for the middle group.

The foreign-born group is more heterogeneous, and it is therefore interesting to look at variables that are important to labor market outcome. A general trend related to all factors is that time consistently reduced the participation rate for the cohort under investigation. Looking at the different age groups, we see about the same patterns as for the Swedish-born but on a higher level. For example, in the initial year the participation rate was four times as large. The number of years in the country is an important variable and we see that those who have been in the country for longer then 22 years have converged to what could be interpreted as a long-run level of around 3%. This is of course a relatively old group of people and they should therefore be compared with the oldest age category, which shows relatively low participation rates as well.

When looking at country of origin we see that there is a distinct difference between those from Europe and those with an origin outside Europe, where the later group has a much larger participation rate. The same applies to those who come to Sweden as refugees.

5 Results

Welfare participation differs greatly among different groups of people and those most exposed are typically young people, single mothers, and immigrants. In this study we separate the analysis between those born in Sweden and those born elsewhere. This is important since it is well-known that the welfare behavior differs greatly between these groups and that the factors affecting their participation behavior are different, as

could be seen in the data section. These differences are believed to be part of the state dependence as well, which will receive special attention in the sections below.

5.1 Swedish-born individuals

Table 3 contains estimates from the dynamic discrete choice model for the Swedishborn group, and is based on a simulated maximum likelihood function using 40 simulated draws per individual and time period. The table shows the estimates of the initial condition equation and the participation equation, and the estimates of the fixed time effects that were estimated as part of the participation equation. The parameters of the initial condition equation are of less interest since its main purpose was to control for the endogenous initial period. The focus will therefore be on the participation equation.

The fixed time effects are all significant and their sizes follow the general time trend in welfare participation that peaked 1997. The overall results are in line with those found in the literature. The effect from continuous age is negative, implying that the likelihood of receiving social assistance decreases with age. This corresponds to the situation that young people more often are exposed to welfare, since they are new on the labor market and not properly established. For each additional age-year, the likelihood of going on welfare decreases by 0.2 percentage points.

It is also well established that years of education is negatively associated with the propensity to end up on welfare, and the results here indicate that an increase in the educational level reduces the risk of going on welfare. The transition from primary schooling to a secondary schooling degree reduces the likelihood by 1.2 percentage points and this figure more than doubles in the transition to a post secondary degree.

Official statistics show that there is a great deal of regional variation in welfare expenditure as well as in the number of participants among municipalities. It has been estimated that around 70 % of the variation in welfare cost among municipalities can conditions be explained by labor market and population structure (Budgetpropositionen, 2005). One would expect that the labor market conditions would be more favorable in city regions, since the supply of jobs is greater there compared to the countryside. However, no such spillover effect from employment opportunities on living in a city region could be found here.

Table 3 Estimation results from a dynamic discrete choice model on social assistance for **Swedish-born people** using a full covariance structure with normalized variance in all time periods (simulated maximum likelihood function using 40 draws)

•	Initial condition		Partic	cipation equ	uation	Fixed time effects		
Observed factors	P.E.	S.E.	P.E.	S.E.	M.E. ⁷	Year	P.E.	
Age/100	-0.599	0.391	-1.103*	0.211	-0.048	1990	-1.694*	
Educational level (CG: Prin	mary school)							
Secondary schooling	-0.383*	0.062	-0.262*	0.035	-0.012	1991	-1.723*	
P. Secondary schooling	-0.867*	0.123	-0.741*	0.058	-0.032	1992	-1.795*	
City region	0.027	0.081	-0.056	0.043	-0.003	1993	-1.719*	
Cohabitant	-0.638*	0.069	-0.436*	0.031	-0.020	1994	-1.841*	
Children, < 6 years	0.348*	0.072	0.128*	0.034	0.006	1995	-1.918*	
Unemployed	0.723*	0.077	0.401*	0.028	0.023	1996	-1.868*	
Regional rate of welfare participation/10	0.961*	0.289	0.839*	0.109	0.033	1997	-1.954*	
SA norm/10K	0.164	0.174	0.258*	0.047	0.012	1998	-1.946*	
Average regional welfare duration/10	-0.268	0.218	-0.384*	0.086	-0.023	1999	-1.941*	
Structural state dependence			0.897*	0.072	0.041			
Log likelihood	-662	7.42						
Sample size	820)50						
Structural state dependence (100 draws)			0.872*	0.068				
Alternative error schemes	Log likelihood		LR-test					
First order Markov	-693	8.76	622.68*					
Component of variance	-667	5.43	95.96*					

Note: P.E. = Parameter estimates; S.E. = Standard errors.; M.E. = Marginal effects. * indicates significance at the 5 % level. LR-test refers to a log likelihood ratio test comparing alternative specifications where the main specification works as base. The critical value at 45 degrees of freedom is 61.65.

When looking at simple correlation measures between city region and welfare participation, one typically receives significant correlation estimates, even though they are small. However, when controlling for unobserved individual differences these effects typically disappear. This could be an indication of a sorting structure which implies that individuals with a higher propensity to end up on welfare tend to stay in city regions.

In the literature it is typically argued that unemployment together with household separations explain the major part of the temporary need for social assistance in some households. One would therefore expect that cohabitation and marital status would reduce the likelihood of living on welfare. This is confirmed by the results, indicating

probability function for period t, where all other time periods have been integrated out. For simplicity reasons, the discrete variables have all been treated as being continuous. However, the continuous treatment is believed to be a good approximation of the discrete counterpart. The derivatives are calculated using a finite difference formula.

⁷ The marginal effects calculated here are based on the full model and represent the mean marginal effects over time and individuals: $\frac{1}{NT} \sum_{i=1}^{N} \sum_{t=1}^{T} \frac{\partial}{\partial x_1} \Phi^*(y_{it} = 1 \mid x)$, with $\Phi^*(y_{it} = 1 \mid x)$ being the marginal

that living together with someone in a household reduces the likelihood of going on welfare by 2 percentage points.

Households with children typically have a strained economic situation, especially when both parents and their children are young, since being young is associated with lower earnings. Having children under age six increases the likelihood of welfare dependence by 0.6 percentage points.

Being unemployed seems to be one obvious reason why some people end up living on welfare. But when the analysis is made on the general population aged 18-50 and related to a random individual from that population, the link is not that strong. This is due to most people depending on unemployment insurance and not social assistance when unemployed. The likelihood of being dependent on welfare when unemployed increases by only 2.3 percentage points, which is by no means the largest effect in the model.

A more interesting effect on individual welfare behavior comes from the local (municipal) average welfare participation rate. This variable stems from the effect of the influence of environmental or local networks on welfare participation. Åslund and Fredriksson (2005) investigated whether the size and the characteristics of ethnic enclaves have any causal effect on welfare use among immigrants. They found that individual welfare use increased by 2.6 percentage points in response to an increase in the share of welfare recipients by 10 percent. This is in line with our study, which also finds a positive relation between the share of welfare recipients and the individual propensity to live on welfare: When the share of welfare recipients increases by 1 percentage point, the propensity increases by 0.3 percentage points.

The size of the social assistance norm mechanically regulates the size of the group of people eligible for social assistance. If the norm is larger, the eligible group become larger, and obviously more people may then choose to live on welfare. However, it is reasonable to believe that the largest effect concerns those on the margin of being a welfare participant, which implies that the overall effect on the population should be quite small. If the yearly norm is increased by 10,000 SEK the propensity to receive social assistance increases by 1.2 percentage points.⁸

Another interesting variable measures whether the local (municipal) average duration on welfare affects the propensity to live on welfare. To be more exact, the measure

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⁸ 10.000 SEK corresponds to 1.052 EUR (August, 2007).

represents the local average number of welfare months during a given year, which should be seen as a proxy for dependency duration, or the strength of the dependency that welfare recipients have in a given municipality. The variable is found to have a negative effect on the propensity to live on welfare. The rationale behind this relationship is not obvious. In the data we find no statistical relation between local welfare duration and unemployment or welfare participation if we look at simple unconditional correlation measures. However, we find a strong and positive statistical relationship between local welfare duration and the local rate of welfare participation (0.28) and large city regions (0.23). This implies that when the local rate of welfare participation and the local average welfare duration are both high, the unconditional effect on welfare participation is cancelled, and when controlling for individual heterogeneity the effect becomes negative. At this point it is still an unanswered question whether this effect is behavioural or spurious.

The last variable in the specification is related to welfare persistence, and the effects of welfare participation over time. That is, when people are introduced to social assistance, a change in their propensity takes place that makes it harder to leave the welfare state, which implies negative duration dependence. In the dynamic literature using continuous duration models, this is a phenomenon that is often noted and investigated. The finding of negative duration dependence is subject to more than one interpretation that differs depending on whether the analysis controls for unobserved heterogeneity. When that is not the case, the duration dependence might be spurious.

The effects of structural state dependence, which is measured using a first order Markov process, constitute the single largest participation effect among those included in the analysis. It implies that if an individual receives welfare in the previous year, he or she then has a 4.1 percentage point increased propensity to receive it in the present year. This has important policy implications since any short-term economic policy measure that increases the participation rate will have long-term consequences that might be difficult to solve, at least in the short-run.

The general error structure was, in a second step, restricted to a specific structure: a first order Markov process and a component of variance structure. Table 3 reports the corresponding log-likelihood values and likelihood-ratio tests. As can be seen, the general structure offers a significant improvement. However, the component of variance structure seems to be a relatively good approximation to the general structure in this case. The general behavior of the coefficient for structural state dependence is

that it is biased upwards, and that the more restrictive the error structure is, the more the bias increases.

Table 4 Estimation results from a dynamic discrete choice model on social assistance for **foreign-born people** using a full covariance structure with normalized variances in all time periods (simulated maximum likelihood function using 40 draws)

in an time perious	Initial condit		Participat			Fixed time effects		
Observed factors	P.E.	S.E.	P.E.	S.E.	M.E.	Year	P.E.	
Age/100	-0.637*	0.281	-0.732*	0.161	-0.093	1990	-0.679*	
Educational level (CG: pr	imary school)							
High school	-0.200*	0.050	-0.128*	0.023	-0.016	1991	-1.117*	
College	-0.706*	0.107	-0.387*	0.033	-0.044	1992	-1.170*	
City region	-0.044	0.049	0.015	0.026	0.005	1993	-1.104*	
Cohabitant	-0.491*	0.047	-0.308*	0.021	-0.041	1994	-1.221*	
Children, < 6 years	0.208*	0.046	0.121*	0.022	0.014	1995	-1.222*	
Unemployed	0.393*	0.055	0.288*	0.018	0.051	1996	-1.132*	
Regional rate of welfare participation/10	0.357*	0.174	0.432*	0.065	0.041	1997	-1.168*	
Welfare norm/10K	-0.076	0.177	0.094*	0.028	0.011	1998	-1.281*	
Average regional welfare duration/10	0.191	0.192	-0.274*	0.081	-0.049	1999	-1.195*	
Country of origin	(CG:Nordic cour	ntries)						
Western Europe	-0.344*	0.095	-0.244*	0.052	-0.030			
Eastern Europe	-0.061	0.106	-0.025	0.055	-0.010			
Southern Europe	-0.488*	0.106	-0.158*	0.053	-0.021			
Middle East	0.296*	0.098	0.379*	0.048	0.036			
Rest of the world	0.295*	0.069	0.262*	0.038	0.031			
	gration (CG: 0-4)	vears)						
5-9 years	-0.505*	0.059	-0.128*	0.027	-0.015			
10-14 years	-0.667*	0.065	-0.231*	0.033	-0.028			
15-22 years	-0.590*	0.063	-0.330*	0.035	-0.039			
>22 years	-0.701*	0.076	-0.441*	0.039	-0.053			
Refugee	0.362*	0.079	0.070*	0.035	0.010			
Structural state			1.041*	0.053	0.125			
dependence Log-likelihood	150	11.20	1.0.1	0.000	0.120			
Sample size		11.28						
Structural state		070						
dependence (100 draws)			1.018*	0.047				
Alternative error schemes	Log-lik	elihood	LR-test					
First order Markov	-162:	54.07	885.58					
Component of variance	-1592		114.53					

Note: P.E. = Parameter estimates; S.E. = Standard errors.; M.E. = Marginal effects. ** indicates significance at the 5 % level. LR-test refers to a log likelihood ratio test comparing alternative specifications where the main specification works as base. The critical value at 45 degrees of freedom is 61.65.

5.2 Foreign-born individuals

We now turn to the second group under investigation in this study, namely the foreign-born group. The results from the simulated maximum likelihood function are presented in Table 4. As with the Swedish-born group, this table contains parameter estimates from the initial condition equation as well as from the main participation equation, which includes fixed time effect dummies. Additionally it contains extra

⁹ Mean marginal effects over time and individuals. See Footnote 7.

observable factors directly related to the foreign-born group, namely country of origin, number of years in the country, and whether the individual came from a refugee country. As before, the discussion will focus on the parameters from the participation equation.

The level of the fixed time effects are much smaller compared to those in the Swedish-born group, and the evolution over time is bimodal, with a first peak in 1994/95 and a second and larger peak in 1998. It is always difficult to interpret intercepts since they are affected by both the variables included and the choice of reference dummy for groups of dummies. However, the general trend is similar to that of the Swedish-born group.

The observed factors in common with the Swedish-born group show about the same effects on welfare propensity when it comes to direction, but there are some differences related to size that are worth mentioning. Continuous age shows a twice as large effect, which means that the welfare behavior differs more among different age groups than for the Swedish-born. Being young and being born in another country are two factors working in the same direction in terms of propensity for welfare participation.

The effects of higher education are at about the same level, while the effect of living in a large city region is almost twice as large, even though it is still very small. Marital status is an important factor, and living together with someone reduces the likelihood by 4 percentage points, which is twice the number for the Swedish-born. We know from the data section that the share of the welfare receiving households with several family members are growing among immigrants. However, this is a phenomenon that appeared in the second half of the 1990s, and we are analyzing and following a random sample taken in 1990, which obviously does not follow this pattern. Hence, the described phenomenon is mainly related to newly arrived immigrants and refugees arriving in the country with their whole families.

A related factor is the presence of younger children in the household. Having children is often associated with an increased risk of living on welfare, and having children younger than age six increases the likelihood by 1.4 percentage points; an effect twice the number of the natives.

Unemployment is a natural cause for welfare dependency, especially for immigrants where it increases the propensity by 4.1 percentage points. The situation is especially difficult when an individual is new in the country, and has an origin outside Europe.

The matching problem on the labor market is related to both structural and individual factors that make it difficult for immigrants to integrate.

The effect of regional rate of welfare participation and the size of the welfare norm is about the same for immigrants as for the Swedish-born. The local average welfare duration on the other hand is much larger though, and as for the Swedes the effect is negative.

Country of origin is important, and the country-groups in the specification are in relation to Nordic-born people, a group very similar in their characteristics to the Swedish-born group. Compared to the Nordic-born, we can identify two groups: one with a larger propensity for welfare and another with a lower propensity. If from Western or Southern of Europe, the propensity is reduced by 2-3 percentage points compared to the Nordic group. People from Eastern Europe have about the same propensity as the Nordics. If from the Middle East or the rest of the world, the effect is an increase in the propensity for welfare by 3-4 percentage points compared to the Nordic group. From these results it is very clear that there is a distinct difference whether a person is from Europe or from a country outside Europe in terms of welfare participation.

The second important immigrant-specific factor for welfare participation is the number of years since immigration. The comparison group consist of those who had been in the country for less then five years. Compared to this group it is clear that the longer the person has been in the country, the more unlikely it is that he or she ends up on welfare. A person who has been in Sweden for more than 22 years has a 5.3 percentage point reduction in propensity, compared to the newly arrived, and this is one of the largest effects in this specification.

During the 1990s, Sweden received a large number of refugees, and many of them stayed in Sweden for many years. This implied a large increase in welfare use, since they came in large numbers and often had problems integrating into the labor market. Our group does not include all these new refugees and therefore the effect is relatively modest, corresponding to a propensity increase by 1 percentage point.

The last measure related to welfare persistence is more interesting. The effect from structural state dependence is very large and three times as large compared to the Swedish-born. This implies that previous experience of welfare increases the propensity by 12.5 percentage points.

As for the Swedish-born, the effect on the fit of the model was tested for different more restrictive error structures. The conclusion is about the same here, with a significant difference between the general error structure and the two alternative error structures, and an increased bias in the coefficient of the structural state dependence that increases when restrictions are imposed on the error structure.

5.3 Persistence

The results presented above show that the structural state dependence in social assistance participation exists, is important, and differs greatly between Swedish-born and foreign-born people. In this section the effect of structural state dependence is decomposed and analyzed with respect to a number of observed factors in order to see how the size of the structural state dependence may change due to changes in those factors. Table 5 presents the parameter estimates for the different factors, and some effects do stands out.

Table 5 Average marginal effects on structural state dependence

	Swedish-born individuals			Foreign-born individuals			
Factors	P.E.	S.E.	M.E. 10	P.E.	S.E.	M.E.	
Constant	0.621*	0.235	-	0.286	0.156	-	
Age/100	0.551	0.352	0.024	0.671*	0.214	0.078	
City region	-0.162*	0.078	-0.007	-0.070	0.041	-0.008	
Cohabitant	0.098	0.067	0.004	0.153*	0.034	0.018	
Unemployed	-0.445*	0.055	-0.020	-0.423*	0.032	-0.049	
Regional rate of welfare	0.311	0.199	0.013	0.205	0.106	0.024	
participation/10							
Welfare norm/10,000	-0.066	0.116	-0.003	0.407*	0.132	0.047	
Average regional welfare duration/10	0.597*	0.189	0.027	0.383*	0.150	0.045	
Country of origin (CG: No	ordic countries	s)					
Westenr Europe				0.150	0.096	0.017	
Eastern Europe				0.177*	0.084	0.021	
Southern Europe				0.152	0.080	0.018	
Middle East				0.138	0.071	0.016	
Rest of the world				-0.029	0.054	-0.003	
Years since immigration (C	CG: 0-5 years)					
5-9 years				-0.061	0.055	-0.007	
10-14 years				-0.152*	0.059	-0.018	
15-22 years				-0.061	0.062	-0.007	
>22 years				-0.006	0.065	-0.001	
Refugee				-0.057	0.053	-0.007	

Note: P.E. = Parameter estimates; S.E. = Standard errors.; M.E. = Marginal effects. * indicates significance at the 5 % level.

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¹⁰ Mean marginal effects over time and individuals. See Footnote 7.

For the Swedish-born there are four significant coefficients, including the constant term. The first significant parameter is related to living in a city region and reduces the size of state dependence. From the earlier discussion we know that living in a city region increases the likelihood of receiving social assistance in general. However, this likelihood is reduced if the person received social assistance in the previous period. This implies that the persistence in social assistance is lower in city regions, even though the probability to receive social assistance is larger in general. One possible explanation for this could be the greater supply of jobs in urban regions, which increases the possibility for households to live on their own earnings.

The second significant parameter refers to unemployment which also has a negative effect on the state dependence. This means that the overall probability to live on welfare when unemployed, is reduced if the household received social assistance in the previous period, which is to say that state dependence is decreasing with the event of being unemployed. Since this analysis is based on a general population we know that most people receive cash assistance or unemployment insurance when being unemployed. We therefore believe that the estimated effect of unemployment on state dependence is contaminated by this general behavior.

The third significant effect for the Swedish-born refers to the effect from the average regional welfare duration. The effect is positive, which means that the persistence is stronger in regions with high average welfare durations, which is an indication that group behavior has an influence on the individual.

The foreign-born group has more factors to consider and therefore more significant effects can be found. As could be seen in the previous sections, marriage and cohabitation is an important factor and is strongly related to receiving social assistance. Apparently it has an important effect on structural state dependence as well. Being unemployed has a negative effect on the size of structural state dependence, and the size of this effect is about the same as for the Swedish-born. Another important factor for the foreign-born as opposed to the Swedish-born is the size of the social assistance norm. If the norm is increasing in the present period, then the size of the structural state dependence is also increasing, and therefore the persistence of social assistance is strengthened. This effect could not be found for the Swedish-born.

Country of origin is another factor that could potentially be of importance for any state dependence. However, only small effects could be found and only the group

from Eastern Europe show a significant increased effect. Somewhat surprisingly we found no increased effect of being from a refugee country.

Table 6 A third order autoregressive specification of structural state dependence

		Swedish-bori	ı	Foreign-born					
Period	P.E.	S.E.	M.E.	P.E.	S.E.	M.E.			
<i>t</i> -1	0.763	0.092	0.028	1.001	0.076	0.099			
<i>t</i> -2	0.450	0.101	0.016	0.253	0.070	0.024			
<i>t</i> -3	0.241	0.099	0.009	0.190	0.079	0.021			

To further analyze the behavior and size of structural state dependence in welfare participation, we include more lags to investigate how many years into the future the experience of social assistance affects the likelihood of receiving social assistance. Table 6 includes the estimated coefficients and the corresponding marginal effects for those coefficients that were significant when using a third order autoregressive specification. It turns out that the number of lags that were significant was the same for both Swedish-born and foreign-born people. However, the initial year effect was more than three times as large for the foreign-born group.

5.3.1 Consequences on earnings and disposable income

In order to investigate the consequences of social assistance participation, we will analyze the pattern and dynamics of household earnings and disposable income. ¹¹ The point of departure is a group of people (households) that received social assistance for the first time in 1990. ¹² This group had no previous experience of welfare for at least three years. In the previous section we saw that the structural state dependence decayed over time and showed no significant effect after three years. We therefore believe that three years is a sufficient number to make the initial year exogenous. We

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¹¹ Since our household definition in SWIP is based on tax register information it is difficult to receive a correct figure for the household incomes. We therefore use earnings and disposable income for the sample person, to represent household income. The disposable income variable available in SWIP is a construction made by Statistics Sweden and represents the individual component (household share), which therefore includes all the transfers to the household.

¹² When constructing the group of new social assistance recipients, we used the same selections as for the main model, with the exception that we used the whole sample of SWIP instead of just a subsample. From this larger sample we selected those who received social assistance in 1990 and eliminated those who received social assistance in any of the three previous years (1987-1989). For the matched comparison group we applied the same criteria of no previous social assistance experience and no social assistance in 1990. The number of Swedish-born newly introduced social assistance recipients was 260, while the number for the foreign-born was 370.

use propensity score matching to find a comparison group with no social assistance.¹³ The matching procedure is used in an attempt to reduce or eliminate any selection bias between the groups. The comparison group will constitute the counterfactual state for those who received social assistance in 1990, and should be interpreted as such.

Figure 6 shows the evolution of the response to receiving welfare in terms of earnings and disposable income for the Swedish-born and their counterfactuals. Once the group has received social assistance, we see that there is a dramatic effect in earnings compared to the case of not being introduced to any social assistance in 1990.¹⁴ Remember that before 1990 there is no significant difference in earnings between the groups while in 1990 and on the difference is large and significant. It is remarkable to see how the differences in earnings persist over time and that there is no sign of any convergence even after ten years. This is an indication of a very strong negative effect with long term consequences on the labor supply from social assistance experience.

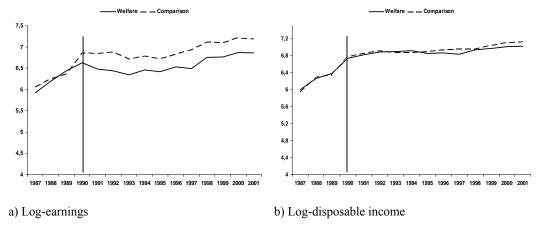


Figure 6 Response to welfare receipt in 1990 for Swedish-born people

When looking at disposable income there is no effect and no significant difference between the groups. Hence, even though earnings decrease following the introduction of social assistance, the disposable income is more or less constant during the same period.¹⁵ This means that the group that was introduced to social assistance in 1990

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¹³ We use the single neighbor matching estimator and use the same factors as those used in the dynamic model, with the exception that we exclude the lagged dependent variables and include previous earnings and disposable income. The propensity score estimates and balancing scores may be received from the author by request.

¹⁴ Earnings is defined as income from work and includes sickness benefits and compensation for parental leave when leave from work is occasional. It does not include unemployment compensations and pensions.

¹⁵ Both earnings and disposable income are deflated and expressed in 1990 money values.

continued to live on transfers from the welfare system for many years thereafter, while the comparison group with significantly higher earnings (between 30 and 40 percent higher) lived on their own earnings to a higher degree.

For the foreign-born the situation is comparable. Figure 7 shows their response to social assistance in 1990, and we can see that it has a strong negative effect on earnings but no effect on disposable income. For the foreign-born, the disposable income is actually significantly higher in the following year (1991) for those who received social assistance. However, this initial effect is only temporary and vanishes in the following years. Overall, the behavior is about the same for the Swedish- and the foreign-born welfare recipients.

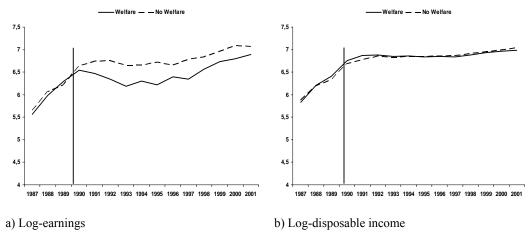


Figure 7 Response to receiving welfare in 1990 for foreign-born people

6 Summary and Conclusions

We estimate the size and the shape of structural state dependence in welfare participation in terms of social assistance for natives and immigrants in Sweden. The effects were estimated using a dynamic discrete choice model controlling for the initial condition and unobserved heterogeneity. Four parts of the structural state dependence were analysed.

The first part focused on the estimated size of the structural state dependence within the framework of a first order Markov process as an aggregated measure. We found that the effect is three times as large for immigrants as it is for natives. Furthermore, among the explanatory variables included in the specification, structural state dependence accounted for the single largest effect for both natives and immigrants with marginal effects of 4.1 and 12.5 percentage points, respectively. Hence, the behavioural response to the experience of social assistance is very strong and significant and has long-term consequences, especially among immigrants.

This aggregated effect from structural state dependence is expected to vary within each group. In the second step of the analysis we therefore disaggregated the effect with respect to a number of explanatory variables we believe could affect state dependence. For the natives, surprisingly few observed factors turned out to be significant. Living in a city region and being unemployed had a negative effect on structural state dependence, while average regional welfare duration showed a positive association.

For the foreign-born, being unemployed was found to be negatively associated with state dependence, while the size of the social assistance norm, the average regional welfare duration, and age had a positive effect. Country of origin and the number of years in the country are two other important determinants for the likelihood of receiving welfare in general. However, it turns out that they had no influence on state dependence among immigrants.

The third part of the analysis relaxed the assumption of a first order Markov process and allowed for more lags. Three lags turned out to be significant for both natives and immigrants, which means that the increased likelihood of returning to social assistance disappears after three years. The difference in effect between natives and immigrants was largest in the first year, and dramatically decreased in size in the second year. While the marginal effect continued to shrink for the natives, the size of the marginal effect was about the same after three years as after two at just above 2 percentage points. The size of the marginal effect reduced to a number slightly below 2 percentage points in the following year, but was not significant. Even though the number of significant years remained the same for natives and immigrants, it seems like the persistence was stronger in the latter group.

This is in accordance with the conclusions of the fourth part of the analysis, which focused on the earnings and disposable income of groups of people introduced to social assistance 1990. The effect on earnings was dramatic, but non-existent in terms of disposable income. The effect on earnings was slightly stronger for the immigrants. The important point to note is the zero effect on disposable income, which clearly indicates that the economic incentives to leave the welfare dependency are very small.

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Appendix

Table A1 Estimated intertemporal covariance matrix for Swedish-born

Time	<i>T</i> =1	<i>t</i> =2	<i>t</i> =3	<i>t</i> =4	<i>t</i> =5	<i>t</i> =6	t=7	<i>t</i> =8	t=9	t=10
t=1	1.0	0.418	0.479	0.466	0.446	0.493	0.411	0.446	0.351	0.399
<i>t</i> =2		1.0	0.409	0.465	0.456	0.452	0.438	0.383	0.397	0.369
<i>t</i> =3			1.0	0.414	0.497	0.523	0.476	0.410	0.412	0.385
<i>t</i> =4				1.0	0.457	0.455	0.527	0.511	0.489	0.452
<i>t</i> =5					1.0	0.476	0.494	0.459	0.510	0.381
<i>t</i> =6						1.0	0.503	0.529	0.525	0.486
t=7							1.0	0.443	0.512	0.459
<i>t</i> =8								1.0	0.456	0.515
t=9									1.0	0.530
<i>t</i> =10										1.0

Note: All parameters are significant at the 1% level

 Table A2 Estimated intertemporal covariance matrix for foreign-born

Time	<i>t</i> =1	<i>t</i> =2	t=3	<i>t</i> =4	<i>t</i> =5	<i>t</i> =6	t=7	t=8	t=9	<i>T</i> =10
t=1	1.0	0.402	0.437	0.396	0.425	0.405	0.383	0.345	0.349	0.273
t=2		1.0	0.361	0.425	0.441	0.442	0.410	0.351	0.385	0.296
<i>t</i> =3			1.0	0.318	0.454	0.445	0.395	0.372	0.374	0.297
<i>t</i> =4				1.0	0.403	0.349	0.476	0.413	0.452	0.325
<i>t</i> =5					1.0	0.402	0.486	0.457	0.518	0.374
<i>t</i> =6						1.0	0.398	0.518	0.471	0.369
t=7							1.0	0.461	0.555	0.434
<i>t</i> =8								1.0	0.473	0.413
t=9									1.0	0.355
<i>t</i> =10										1.0

Note: All parameters are significant at the 1% level