

***Latent Transition Analysis  
of Substance Use Among  
Adolescents in the  
National Longitudinal  
Survey of Youth***

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## **Abstract**

The present study examines gender differences in substance use onset among 886 adolescents surveyed in the National Longitudinal Survey of Youth (NLSY). Three waves of data on adolescent substance use were analyzed using Latent Transition Analysis (LTA). To test the gateway hypothesis of substance use (Kandel & Yamaguchi, 1993; Collins, 1998a), transition probabilities of adolescent exposure to three substances, alcohol, cigarettes and marijuana, were estimated. In addition, a strategy similar to that employed by Collins, Graham, Long and Hansen (1994) was used to test gender differences in substance use onset among the NLSY adolescents.

Nested models with different degrees of parameter restrictions were fitted to the data. First, a model that represents gender differences in all three sets of parameters (Big Rho's, Delta's, & Tau's) was fitted. Other models that constrained some of these parameters to be equal between gender were fitted subsequently. Model fit statistics were compared among models and the best fitting one was selected for interpretation. The results suggested that the model that represented no gender differences in all three parameter sets fit the data best. This suggests there are no gender differences in the types of substance use patterns (number and types of latent statuses), the proportion of adolescents falling into each latent status and the transition pattern of substance use. In addition, the study supports the gateway hypothesis that cigarette and alcohol usage are precursors to the use of marijuana.

Due to the high percentage of non-response to substance use items in the survey as well as the weak measure of marijuana use, the present findings have to be interpreted with caution.

## **1. Introduction**

The present study examines gender differences in substance use onset among 886 adolescents surveyed in the National Longitudinal Survey of Youth (NLSY). Three waves of data on adolescent substance use were analyzed using Latent Transition Analysis (LTA). To test the gateway hypothesis of substance use (Kandel & Yamaguchi, 1993; Collins, 1998a), transition probabilities of adolescent exposure to three substances, alcohol, cigarettes and marijuana, were estimated. In addition, a strategy similar to that employed by Collins, Graham, Long and Hansen (1994) was used to test gender differences in substance use onset among the NLSY adolescents.

## **2. Sample**

Our sample consists of 886 adolescents participating in the NLSY, which is a national survey conducted by the Bureau of Labor Statistics, U.S. Department of Labor. The mothers of these adolescents were initially enrolled in the study in 1979, using a standard multi-stage, area probability sampling method. Data were collected annually on these young women's education, job training and work experience. The sampling scheme also included an oversampling of Hispanics, African Americans, and disadvantaged non-Hispanic, non-African American youths.

Starting from 1986, offspring of the NLSY young women were surveyed biennially. Our sample consists of these offspring of the NLSY, whose ages were 10 to 14 in 1988, the time when the first wave of data used in our study were collected. One of the reasons for our choice of sample is that in the survey only children aged 10 and older were asked substance-use-related questions. Furthermore, by choosing such an age range, we are

restricting our sample to children entering or studying in the middle school at the time of our first wave of measurement. So, children in our sample are expected to be exposed to similar school environments. Our analysis included data from two more waves of data collected in 1990 and 1992. Out of the 886 adolescents, 444 were males; 147 were Hispanic, 355 were Black and 384 were non-Hispanic, non-Black. Many of the mothers of these children gave birth to their first child in their teen years.

There is a relatively high proportion of non-response in the data. The proportion of missing information for the smoking indicator increased from 8.2% to 48% across the three waves of measurement. Similarly, the proportion missing for the drinking indicator increased from 8.4% to 48.1%. Likewise, the proportion missing for the marijuana use indicator increased from 8.4% to 48.1%.

### **3. Measures**

General information on the children's development was collected through standardized assessments and parent interviews. Those children age 10 years and over were also asked to complete questionnaires that measure their family relationships and school experiences. Several items related to substance use were also included in the questionnaire.

Substance use indicators were drawn from the NLSY's Child Self-Administered Supplement (CSAS). (This is a confidential self-report survey given to all children 10 years and older. The procedure was started in 1988. From 1994 onwards, instead of the CSAS, children older than 15 were given the Young Adult Survey (YAS)). Adolescents were asked about their relationships with their parents, the ways they spent their time and their exposure to various substances.

Our study focused on exposure to alcohol, cigarettes and marijuana. Three pairs of items on each of the substances were used in our study.

1. Have you ever smoked a cigarette? (Yes = 1; No = 0)
2. If you have ever smoked, have you smoked a cigarette in the past three (3) months?  
(Yes = 1; No = 0; Never smoked = 2)
3. Have you ever drunk alcohol, other than just a sip or two? (Yes = 1; No = 0)
4. If you have ever drunk alcohol, have you drunk alcohol in the past three (3) months?  
(Yes = 1; No = 0; Have never drunk alcohol = 2)
5. Have you ever used marijuana? (Yes = 1; No = 0)
6. If you have ever used marijuana, have you used marijuana in the past three (3) months? (Yes = 1; No = 0; Never used marijuana = 2)

The items were recoded to form binary items that indicate whether the adolescents had been exposed to a substance (No = 1; Yes = 2; and Missing = 0). The two items were then combined to form a single indicator of substance exposure. Whenever an individual answered positively to one of the items, it was counted as a 'yes' to the single indicator.

#### **4. Analytic Strategies**

Latent Transition Analysis (LTA) (Collins & Wugalter, 1992; Graham, Collins, Wugalter, Chung, & Hansen, 1991; Langeheine & van de Pol, 1991; Collins, Graham, Rousculp, & Hansen, 1997; Collins, Schafer, Hyatt, & Flaherty, in preparation) was used to analyze our data. The statistical procedure is a special case of Latent Class Analysis where the latent variable is dynamic, i.e. changing in systematic ways over time (Collins & Cliff, 1990). LTA is a type of latent Markov model (van de Pol & Langeheine, 1989). It is designed for modeling changes in stage-sequential dynamic latent variables that have been measured in a longitudinal panel design. We are interested in studying the changes

in individuals' substance use patterns, which are called latent statuses in our model, over the six years from 1988 to 1992.

In LTA that only has the dynamic part of the model, three types of parameters are estimated. The first set of parameters are the “big”  $\rho$ 's. These parameters have the same meaning in LTA models as they do in LCA; that is, they represent the probability of a particular item response, conditional on latent status membership. Examining these parameter estimates helps the user to interpret the characteristics of each latent status. The second set of parameters are the  $\delta$ 's, which are estimates of the proportion of the population in each latent status at each occasion of measurement. The third set of parameter estimates are the  $\tau$ 's, or transition probabilities. For a first-order model, like the one in the present study, these are estimates of the probability of moving to a particular latent status in Time 2 conditional on latent status membership at Time 1. In a model that has both static and dynamic elements, two more sets of parameters are estimated. The first set is called “little”  $\rho$ 's. They serve a similar function as the big  $\rho$ 's in the dynamic part of the model and represent the probability of a particular item response, conditional on latent class membership. The second set of parameters in the static part of the model is the  $\gamma$ 's, which represent the estimated proportions of each latent class (for a more detailed description of the LCA and LTA models, please refer to the WinLTA User Guide (Collins et al., 1999)).

WinLTA 2.1 was used to fit our latent transition models (Collins, 1998b). The program, using the EM algorithm in its estimation of model parameters, has the ability to handle missing data.

#### *4.1 Exploratory Search for the Basic Structure of LTA Models for the NLSY data*

Although the data we used finally in our project were those from the Child Self-Administered Supplement Survey (CSAS) collected in years 1988, 1990 and 1992, we started our project fitting LTA models to the 1994 and 1996 NLSY Young Adult Survey (YAS) data . The YAS survey was started in 1994 for children who were fifteen and older. In the YAS, more elaborate questions were asked about substance use, including questions on binge drinking. Because initially we were interested in the relationship between binge drinking and other substance use, we started with the YAS data. To our disappointment, the binge drinking variables contain a large amount of missing values and the models we fitted had very poor fits. After months of trying, we decided to drop the binge drinking variables and switched to the CSAS survey, even though it contains much less information on substance use when compared to the YAS survey. We mention this initial round of model fitting, even though the results were not very promising, because we got a sense of the structure of the LTA models from the analyses. For the binge drinking model, we found that the "best" fitting model we can get among the models we fitted was one with 7 latent classes: "No Use", "Alcohol Only", "Cigarettes Only", "Alcohol and Cigarettes", "Alcohol and Marijuana", "Alcohol, Cigarettes and Marijuana" and "Alcohol and Binge Drinking" ( $G^2 = 1099.77$ , with  $df=161$ ).

In the next stage of analyses, we dropped the binge drinking variables from our models. Based on the results of our previous analyses, we decided to try models with 6 latent classes (basically we eliminated the latent class that related to binge drinking) and to use only one indicator of substance use per substance. Although we have two

indicators per substance in our data, we decided to go with one to see if the model fit better with simpler response patterns. The model fit for these group of models were still poor. The "smallest" model fit statistics we could get was a  $G^2$  of 400.032 with degrees of freedom of 32.

After some quite disappointing experience with the YAS drug use data, we decided to switch to the CSAS data. Furthermore, we decided to combine the two indicators for each drug into a new single indicator. We fitted new models to the drug use items in 1988 and 1990. The fits for these groups of models were generally better than the previous models we fitted for the YAS data. The model with the smallest fit statistics had 5 latent classes: "No use", "Cigarettes Only", "Alcohol Only", "Alcohol and Cigarettes", and "Alcohol, Cigarettes and Marijuana". In this model, equality constraints were placed on  $\rho$ 's across the two times and some  $\rho$ 's for different substances within each time were constrained to be equal ( $G^2 = 93.86$  with  $df=19$ ).

Because we suspected that the amount of missing values decreased the overall fit of the model, we also tried fitted models to listwise-deleted data (after the deletion, the sample size was reduced from 886 to 484). Similar models as the previous ones were fitted and it seemed that the models fit better. The model with the "best" fit statistics has a  $G^2$  of 31.704 with  $df = 31$ ). This model also has equality constraints on the  $\rho$ 's across time. The  $\rho$ 's for some response categories were constrained to be equal.

Next, we fit a model to all three waves of data, the 1988, 1990 and 1992 substance use data in the CSAS survey. We first fit models to the listwise-deleted sample ( $N=378$ ) and then to a sample with missing values ( $N = 886$ ). For the listwise-deleted sample, the "best" fitting model has a  $G^2$  of 207.997 with 460 degrees of freedom. For the



sample with missing values, the "best" fitting model has a  $G^2$  of 598.326 with 481 degrees of freedom.

Throughout our exploratory analyses, we fixed the number of latent classes and the pattern of constraints on the  $\rho$ 's to be the same across the times of measurement. The other parameters in the model, the  $\delta$ 's and the  $\tau$ 's, were freely estimated. On the other hand, we tried different patterns of constraints for the  $\rho$ 's within each time. The "final" model we got from the exploratory analysis had the following pattern of  $\rho$  constraints for all three time-points (See Table 1).

The proportions of adolescents responding with a "yes" to the smoking indicator were hypothesized to be equally high in the "smoking only", "drinking and smoking" and "Using all substance" latent classes. On the other hand, the corresponding proportion were hypothesized to be equally low in the "no use" and the "alcohol only" latent classes. A similar pattern of constraints was imposed for the drinking indicator, i.e.  $\rho$ 's for "no use" and "smoking only" latent classes were set to be equally low and  $\rho$ 's for the rest of the latent classes were set to be equally high.

For the marijuana indicator, a slightly different pattern of constraints was used. The proportion of adolescents responding "yes" to the indicator was hypothesized to be equally low for the "no use", "smoking only" and "drinking only" latent classes. It is hypothesized that the latent class of "drinking and smoking" has a low proportion too but it was allowed to be different from the other latent classes. The proportion in the "all" latent class was expected to be high and different from the other classes.

#### *4.2 Gender as a Latent Class Variable in the Static Part of the LTA Model*

In order to model gender differences in substance use onset, our LTA model has both static and dynamic components. The grouping variable, gender, is introduced into our LTA model as a categorical variable with two “latent” classes. Related to this “latent” variable there is only one manifest item: individual’s gender reported in the survey. The latter has two response categories: male and female. We fixed the  $\rho$ ’s of the ‘male’ response category conditional on the first and the second latent class to be 1 and 0, respectively. At the same time, we fixed the  $\rho$ ’s of the ‘female’ response category conditional on the first and the second latent class to be 0 and 1. This avoided identification problems, and also treated gender as an observed rather than latent variable. By having the grouping variable ‘gender’ in our model, we can test for any differences in the dynamic part of our LTA model with respect to gender (see later section on ‘sequential testing of gender difference model’). All the gender difference models have the following basic structure:

- ❖ 2 “latent” classes: “Male” and “Female”
- ❖ 1 “manifest” indicator (Male vs. Female) with two response categories (yes vs. no)
- ❖ 3 times of measurement: 1988, 1990 and 1992
- ❖ 5 latent statuses: (No Use, Smoking Only, Drinking Only, Smoking and Drinking, All)
- ❖ 3 manifest indicators (Smoking, Alcohol, Marijuana) of the dynamic latent variable at each occasion, each with two response categories (yes vs. no).

The same kind of constraints were imposed on the  $\rho$ 's,  $\delta$ 's, and  $\tau$ 's as in the "final" model we got from the exploratory analyses, i.e. the  $\delta$ 's and  $\tau$ 's were allowed to be

different for different latent classes. The  $\rho$ 's have the same pattern of constraints as in the "final" model. On the other hand, as described below, different models have different equality constraints placed across gender for each group of parameters.

### 5. Sequential Testing of Gender Difference Models

In order to perform a formal test on the gender difference hypothesis, four models with different degrees of constraints on the different groups of parameters were fitted sequentially to the data. Among these four models, Model 4 is the most restrictive and it has the least number of freely-estimated parameters.

- **Model 1:** All  $\delta$ 's and  $\tau$ 's estimated freely; big  $\rho$ 's allowed to vary across genders.
- **Model 2:** All  $\delta$ 's and  $\tau$ 's estimated freely; big  $\rho$ 's constrained equal across genders.
- **Model 3:** All  $\tau$ 's estimated freely;  $\delta$ 's and big  $\rho$ 's constrained equal across genders.
- **Model 4:** All  $\delta$ 's,  $\tau$ 's and big  $\rho$ 's constrained equal across genders.

### 6. Results

All four models have reasonable fit to the data. The  $G^2$ 's are all less than their corresponding degrees of freedom (See Table 2).

Model 4 is the most parsimonious among the four models (see Table 2). Other models do not provide statistically significant improvement in the fit statistics, which is represented by  $\Delta G^2$ , the decrease in fit statistics from one model to the other. When two models are nested,  $\Delta G^2$  is distributed as  $\chi^2$  and can be used to test whether one model is significantly different from a bigger model. Such a result indicates that there are no gender differences in either the substance use pattern or the changes in such a pattern across time.

The measurement quality of the three indicators of substance use is fairly good in the first four latent statuses but the measurement of marijuana use in the fifth latent status is rather weak. Within such a latent class, the estimated probability that an individual has been exposed to marijuana is close to 0.5 for both the ‘yes’ and ‘no’ response categories (see Table 3). Such a weak  $\rho$  indicates that according to our model, nearly half of the people responding ‘no’ to the ‘marijuana’ item were actually exposed to marijuana while nearly half of the people responding ‘yes’ were not exposed. While it is straightforward to interpret the meaning of a latent status that has strong  $\rho$ 's, it is more difficult to label one that has weak  $\rho$ 's in some of the items (see Table 3 for pattern of  $\rho$ 's and labels for each of the latent statuses).

The estimates of the  $\delta$  parameters gave us the estimated proportion of adolescent substance users in each latent class. In 1988, most of the adolescents have not used any substances. The proportion of the “No Use” group fell below 0.5 in 1992. The group that has tried both alcohol and cigarettes and the group that has tried all three substances increased significantly over the six years. Together the two groups comprise about half of the sample in 1992 (see Figure 2).

When we examine the transition matrix, the matrix of  $\tau$  parameters which represent the movement of individuals among the latent classes across time, we found that there was almost no movement of adolescents among the latent statuses between 1988 and 1990, except that about 20% of the adolescents in the “No Use” latent status were exposed to both alcohol and cigarettes in 1990. During the transition from 1990 to 1992, more movement among the latent statuses was seen. About 24% of adolescents in the “No Use” latent status in 1990 moved into the “All” latent status. Almost all the

adolescents in the “Smoke Only” latent status in 1990 moved into other groups. 87% of this group tried alcohol as well as cigarettes. About 20% of the “Drink Only” latent status in 1990 reported having tried cigarettes and marijuana by 1992. However, only about 3% of the “Drink and Smoke” latent status in 1990 said they also tried marijuana in 1992 (See Table 4).

In this version of WinLTA, we do not have a way to test the statistical significance of individual parameters, which is something we want. In a future version of WinLTA (currently being tested), such an option will be available.

## **7. Conclusions**

Latent transition analysis is a promising approach in drug prevention research. It can handle categorical outcome variables and provides estimates of the prevalence and patterns of substance use as well as the transition probabilities among stages of substance use. It provides a means to test theories on substance use. It is also possible to use multiple group analysis to test group differences.

No gender difference was found in the substance use patterns. Such a result is consistent with previous findings (Collins et al., 1994). Whereas Collins et al.’s sample is based on students residing in one state, the present study is based on a national sample. It further supports the belief that similar substance use prevention programs can be applied to both genders.

The gateway hypothesis of substance use is supported by our findings. In our search for the appropriate latent class structure for drug use responses among NLSY adolescents, we did not find a ‘marijuana use only’ latent statuses, while there are both ‘smoking only’ and ‘drinking only’ latent statuses, as well as a latent class that

corresponds to exposure to all three substances. Such a result indicates that both legal substances, cigarettes and alcohol, act as gateway substances that increase the propensity of adolescents' use of marijuana.

The percentage of missing information in the part of NLSY data we used is high. Even though WinLTA can handle missing data, we have to interpret our results with great caution.

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Table 1: Constraints imposed on  $\rho$ 's for the gender difference models

	SMOKING	DRINKING	MARIJUANA USE
NO USE	2	4	6
SMOKING ONLY	3	4	6
DRINKING ONLY	2	5	6
DRINKING AND SMOKING	3	5	7
ALL	3	5	8

Note: The same pattern of constraints was placed on the "yes" and "no" response categories. All cells with the same number have  $\rho$ 's constrained to be equal.

Table 2. Comparison of model fit for the various gender difference models

	<b>G<sup>2</sup></b>	<b>df</b>	<b>ΔG<sup>2</sup><sup>a</sup></b>	<b>Δdf<sup>a</sup></b>
<b>Model 1:</b> All Delta's and Tau's estimated freely; big Rho's allowed to vary across genders	831.093	960	32.434	31
<b>Model 2:</b> All Delta's and Tau's estimated freely; big Rho's constrained equal across genders	847.256	967	16.271	24
<b>Model 3:</b> All Tau's estimated freely; Delta's and big Rho's constrained equal across genders	852.239	971	11.288	21
<b>Model 4:</b> All Delta's, Tau's and big Rho's constrained equal across genders	863.527	991	Baseline	--

<sup>a</sup> ΔG<sup>2</sup> and Δdf were calculated by comparing a model to the baseline model

Table 3. Measurement quality of substance use indicators in the gender difference models

**Big Rho's**

	<b>Smoking (No)</b>	<b>Drinking (No)</b>	<b>Marijuana Use (No)</b>
<b>NOUSE</b>	0.897	0.849	0.981
<b>SMONLY</b>	0.246	0.849	0.981
<b>DRKONLY</b>	0.897	0.121	0.981
<b>DRKSM</b>	0.246	0.121	0.718
<b>ALL</b>	0.246	0.121	0.543
	<b>Smoking (Yes)</b>	<b>Drinking (Yes)</b>	<b>Marijuana Use (Yes)</b>
<b>NOUSE</b>	0.103	0.151	0.019
<b>SMONLY</b>	0.754	0.151	0.019
<b>DRKONLY</b>	0.103	0.879	0.019
<b>DRKSM</b>	0.754	0.879	0.282
<b>ALL</b>	0.754	0.879	0.457

Table 4. Transitions between latent statuses across measurement occasions

Tau matrix for transitions from 1988 to 1990

	<b>No Use</b>	<b>Smoke Only</b>	<b>Drink Only</b>	<b>Drink and Smoke</b>	<b>All</b>
<b>No Use</b>	0.744	0.016	0.027	0.209	0.005
<b>Smoke Only</b>	0.000	1.000	0.000	0.000	0.000
<b>Drink Only</b>	0.000	0.000	0.911	0.000	0.089
<b>Drink and Smoke</b>	0.000	0.000	0.000	1.000	0.000
<b>All</b>	0.000	0.000	0.000	0.000	1.000

Tau matrix for transitions from 1990 to 1992

	<b>No Use</b>	<b>Smoke Only</b>	<b>Drink Only</b>	<b>Drink and Smoke</b>	<b>All</b>
<b>No Use</b>	0.651	0.014	0.100	0.000	0.235
<b>Smoke Only</b>	0.000	0.023	0.104	0.874	0.000
<b>Drink Only</b>	0.000	0.000	0.808	0.000	0.192
<b>Drink and Smoke</b>	0.000	0.000	0.000	0.971	0.029
<b>All</b>	0.000	0.000	0.000	0.000	1.000

Figure 1. Model of Substance Use Transitions

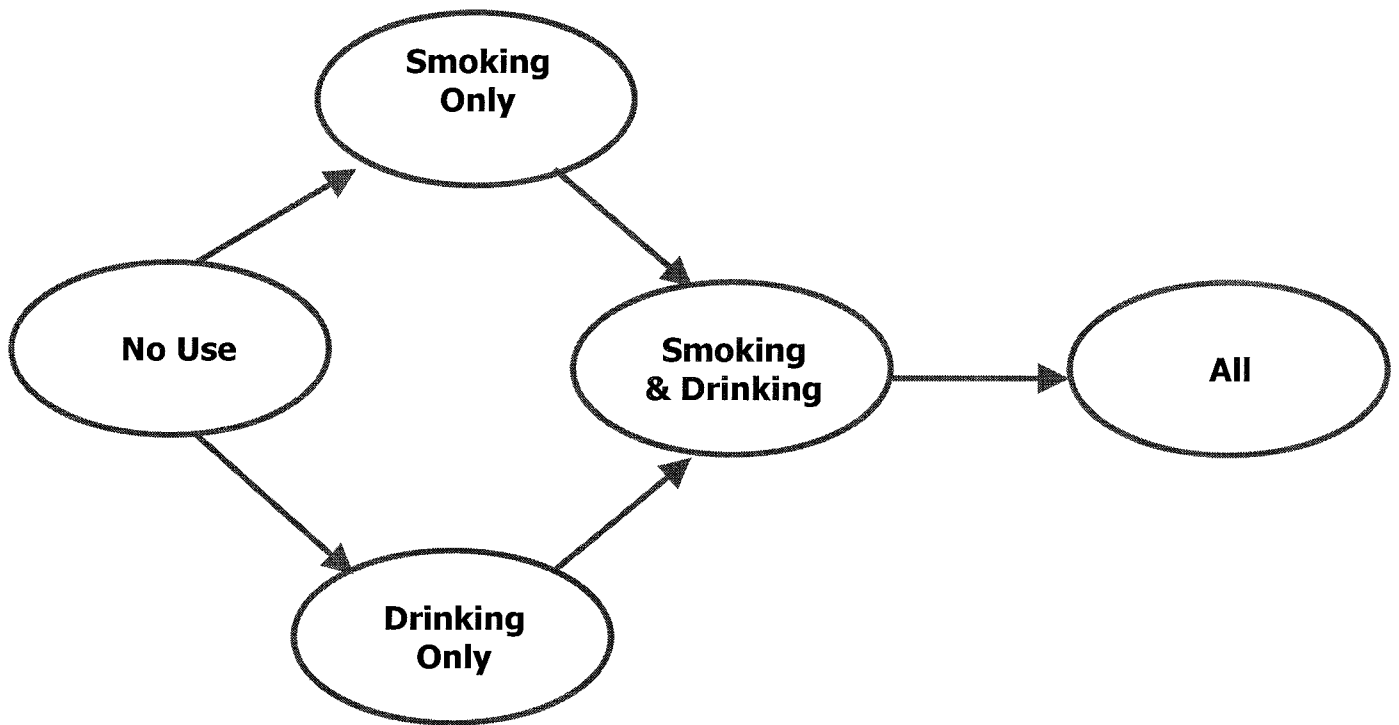
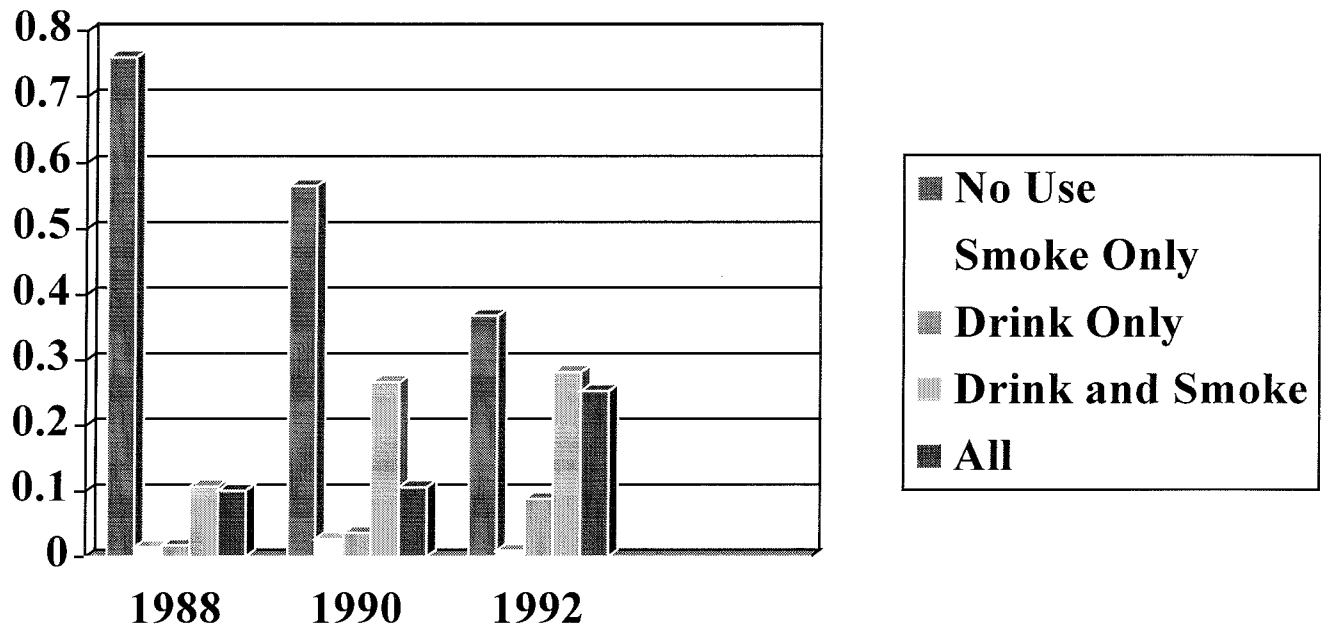


Figure 2. Estimated Proportions of Substance Use Latent Statuses



## Appendix 1. LTA program output for final selected model

PROGRAM STARTED: Sat Jun 12 21:35:00 1999

Model 4B: Same equivalent sets of Tau's, Delta's and big Rho's for males and females

\*\*\*\*\*

INFORMATION ABOUT THIS JOB:

CONTROL DATA SET:  
C:\nspr99\gend3wv\mgcmb311a.cnt

DATA ANALYZED IN THIS RUN READ FROM:  
c:\nspr99\gend3wv\msgcmb3.rsp

OUTPUT PRINTED TO:  
c:\nspr99\gend3wv\mgcmb311a.out

PARAMETER ESTIMATES SAVED TO FILE:  
c:\nspr99\gend3wv\mgcmb311a.prm

STATIC LATENT VARIABLE	YES
NUMBER OF LATENT CLASSES	2
NUMBER OF MANIFEST ITEMS	1

DYNAMIC LATENT VARIABLE	YES
NUMBER OF LATENT STATUSES	5
NUMBER OF OCCASIONS OF MEASUREMENT	3
NUMBER OF MANIFEST ITEMS PER OCCASION	3
TYPE OF PROCESS	FIRST-ORDER

NUMBER OF SUBJECTS	886
NUMBER OF UNIQUE RESPONSE PATTERNS	277
MAXIMUM NUMBER OF ITERATIONS	10000
CONVERGENCE CRITERION	.00000100000000
MISSING DATA IN RESPONSE PATTERNS	YES

PRINT RESIDUALS YES

\*\*\*\*\*

THE FOLLOWING CONSTRAINTS HAVE BEEN SPECIFIED  
WHERE 0=FIXED TO START VALUE

1=FREE

2 OR GREATER MEANS CONSTRAINED EQUAL TO ANY OTHER  
PARAMETER WITH THE SAME DESIGNATION

\*\*\*\*\*

LITTLE RHO PARAMETERS  
 LITTLE RHOS ARE PROBABILITIES OF RESPONSES  
 TO ITEMS MEASURING THE STATIC LATENT VARIABLE  
 CONDITIONAL ON LATENT CLASS MEMBERSHIP

RESPONSE CATEGORY 1

```
| G M |
| E A |
| N L |
| D E |
| E   |
| R   |
|     |
|     |
```

Male 0  
 Female 0

RESPONSE CATEGORY 2

```
| G F |
| E E |
| N M |
| D A |
| E L |
| R E |
|     |
|     |
```

Male 0  
 Female 0

\*\*\*\*\*

BIG RHO PARAMETER CONSTRAINTS  
 BIG RHOS ARE PROBABILITIES OF RESPONSE  
 TO ITEMS MEASURING THE DYNAMIC LATENT VARIABLE  
 CONDITIONAL ON LATENT STATUS, LATENT CLASS, AND TIME

BIG RHO CONSTRAINTS FOR LATENT CLASS "Male" AT TIME 1

RESPONSE CATEGORY 1

```
| S N | D N | M N |
| M O | R O | A O |
| O   | I   | R   |
| K   | N   | I   |
| E   | K   | J   |
|     |     |     |
|     |     |     |
|     |     |     |
```

NOUSE	2	4	6
SMONLY	3	4	6
ALONLY	2	5	6
ALCSM	3	5	7
ALL	3	5	8

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	12	14	16
SMONLY	13	14	16
ALONLY	12	15	16
ALCSM	13	15	17
ALL	13	15	18

BIG RHO CONSTRAINTS FOR LATENT CLASS "Male" AT TIME 2

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	2	4	6
SMONLY	3	4	6
ALONLY	2	5	6
ALCSM	3	5	7
ALL	3	5	8

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	12	14	16
SMONLY	13	14	16
ALONLY	12	15	16
ALCSM	13	15	17
ALL	13	15	18



BIG RHO CONSTRAINTS FOR LATENT CLASS "Male " AT TIME 3

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	2	4	6
SMONLY	3	4	6
ALCONLY	2	5	6
ALCSM	3	5	7
ALL	3	5	8

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	12	14	16
SMONLY	13	14	16
ALCONLY	12	15	16
ALCSM	13	15	17
ALL	13	15	18

BIG RHO CONSTRAINTS FOR LATENT CLASS "Female " AT TIME 1

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	2	4	6
SMONLY	3	4	6
ALCONLY	2	5	6
ALCSM	3	5	7
ALL	3	5	8

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	12	14	16
SMONLY	13	14	16
ALONLY	12	15	16
ALCSM	13	15	17
ALL	13	15	18

BIG RHO CONSTRAINTS FOR LATENT CLASS "Female " AT TIME 2

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	2	4	6
SMONLY	3	4	6
ALONLY	2	5	6
ALCSM	3	5	7
ALL	3	5	8

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	12	14	16
SMONLY	13	14	16
ALONLY	12	15	16
ALCSM	13	15	17
ALL	13	15	18

BIG RHO CONSTRAINTS FOR LATENT CLASS "Female " AT TIME 3

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	2	4	6
SMONLY	3	4	6
ALONLY	2	5	6
ALCSM	3	5	7
ALL	3	5	8

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	12	14	16
SMONLY	13	14	16
ALONLY	12	15	16
ALCSM	13	15	17
ALL	13	15	18

\*\*\*\*\*

CONSTRAINTS FOR GAMMA PARAMETERS

GAMMAS ARE UNCONDITIONAL PROBABILITIES OF MEMBERSHIP  
IN EACH LATENT CLASS OF THE STATIC LATENT VARIABLE

Male 1  
Female 1

\*\*\*\*\*

CONSTRAINTS FOR DELTA PARAMETERS

DELTAS ARE PROBABILITIES OF LATENT STATUS MEMBERSHIP  
CONDITIONAL ON LATENT CLASS

DELTA PARAMETER CONSTRAINTS FOR LATENT CLASS "Male "

TIME 1  
NOUSE 101  
SMONLY 102  
ALONLY 103  
ALCSM 104  
ALL 105

DELTA PARAMETER CONSTRAINTS FOR LATENT CLASS "Female "

TIME 1  
 NOUSE 101  
 SMONLY 102  
 ALONLY 103  
 ALCSM 104  
 ALL 105

\*\*\*\*\*

CONSTRAINTS FOR TAU PARAMETERS  
 TAUS ARE PROBABILITIES OF LATENT STATUS MEMBERSHIP AT TIME T+1  
 (COLUMNS)  
 CONDITIONAL ON LATENT STATUS MEMBERSHIP AT TIME T (ROWS)  
 AND ON LATENT CLASS MEMBERSHIP

TRANSITION PROBABILITIES FOR LATENT CLASS "Male "

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 1  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	201	202	203	204	205
SMONLY	0	206	207	208	209
ALONLY	0	0	210	211	212
ALCSM	0	0	0	213	214
ALL	0	0	0	0	0

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 3

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	301	302	303	304	305
SMONLY	0	306	307	308	309
ALONLY	0	0	310	311	312
ALCSM	0	0	0	313	314
ALL	0	0	0	0	0

TRANSITION PROBABILITIES FOR LATENT CLASS "Female "

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 1  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	201	202	203	204	205
SMONLY	0	206	207	208	209
ALONLY	0	0	210	211	212
ALCSM	0	0	0	213	214
ALL	0	0	0	0	0

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 3

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	301	302	303	304	305
SMONLY	0	306	307	308	309
ALONLY	0	0	310	311	312
ALCSM	0	0	0	313	314
ALL	0	0	0	0	0

\*\*\*\*\*  
 START VALUES  
 \*\*\*\*\*

LITTLE RHO PARAMETERS  
 LITTLE RHOS ARE PROBABILITIES OF RESPONSES  
 TO ITEMS MEASURING THE STATIC LATENT VARIABLE  
 CONDITIONAL ON LATENT CLASS MEMBERSHIP

RESPONSE CATEGORY 1

	G	M
	E	A
	N	L
	D	E
	E	
	R	
Male	1.000	
Female	0.000	

RESPONSE CATEGORY 2

```

| G F |
| E E |
| N M |
| D A |
| E L |
| R E |
|     |
|     |

```

Male 0.000  
 Female 1.000

\*\*\*\*\*

BIG RHO PARAMETERS  
 BIG RHOS ARE PROBABILITIES OF RESPONSES  
 TO ITEMS MEASURING THE DYNAMIC LATENT VARIABLE  
 CONDITIONAL ON LATENT STATUS, LATENT CLASS, AND TIME

RHO PARAMETERS FOR LATENT CLASS "Male " AT TIME 1

RESPONSE CATEGORY 1

```

| S N | D N | M N |
| M O | R O | A O |
| O   | I   | R   |
| K   | N   | I   |
| E   | K   | J   |
|     |     |     |
|     |     |     |

```

NOUSE	0.800	0.800	0.800
SMONLY	0.200	0.800	0.800
ALONLY	0.800	0.200	0.800
ALCSM	0.200	0.200	0.800
ALL	0.200	0.200	0.600

RESPONSE CATEGORY 2

```

| S Y | D Y | M Y |
| M E | R E | A E |
| O S | I S | R S |
| K   | N   | I   |
| E   | K   | J   |
|     |     |     |
|     |     |     |

```

NOUSE	0.200	0.200	0.200
SMONLY	0.800	0.200	0.200
ALONLY	0.200	0.800	0.200
ALCSM	0.800	0.800	0.200
ALL	0.800	0.800	0.400

RHO PARAMETERS FOR LATENT CLASS "Male " AT TIME 2

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.800	0.800	0.800
SMONLY	0.200	0.800	0.800
ALONLY	0.800	0.200	0.800
ALCSM	0.200	0.200	0.800
ALL	0.200	0.200	0.600

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.200	0.200	0.200
SMONLY	0.800	0.200	0.200
ALONLY	0.200	0.800	0.200
ALCSM	0.800	0.800	0.200
ALL	0.800	0.800	0.400

RHO PARAMETERS FOR LATENT CLASS "Male " AT TIME 3

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.800	0.800	0.800
SMONLY	0.200	0.800	0.800
ALONLY	0.800	0.200	0.800
ALCSM	0.200	0.200	0.800
ALL	0.200	0.200	0.600

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.200	0.200	0.200
SMONLY	0.800	0.200	0.200
ALONLY	0.200	0.800	0.200
ALCSM	0.800	0.800	0.200
ALL	0.800	0.800	0.400

RHO PARAMETERS FOR LATENT CLASS "Female " AT TIME 1

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.800	0.800	0.800
SMONLY	0.200	0.800	0.800
ALONLY	0.800	0.200	0.800
ALCSM	0.200	0.200	0.800
ALL	0.200	0.200	0.600

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.200	0.200	0.200
SMONLY	0.800	0.200	0.200
ALONLY	0.200	0.800	0.200
ALCSM	0.800	0.800	0.200
ALL	0.800	0.800	0.400



RHO PARAMETERS FOR LATENT CLASS "Female " AT TIME 2

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.800	0.800	0.800
SMONLY	0.200	0.800	0.800
ALONLY	0.800	0.200	0.800
ALCSM	0.200	0.200	0.800
ALL	0.200	0.200	0.600

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.200	0.200	0.200
SMONLY	0.800	0.200	0.200
ALONLY	0.200	0.800	0.200
ALCSM	0.800	0.800	0.200
ALL	0.800	0.800	0.400

RHO PARAMETERS FOR LATENT CLASS "Female " AT TIME 3

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.800	0.800	0.800
SMONLY	0.200	0.800	0.800
ALONLY	0.800	0.200	0.800
ALCSM	0.200	0.200	0.800
ALL	0.200	0.200	0.600

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.200	0.200	0.200
SMONLY	0.800	0.200	0.200
ALONLY	0.200	0.800	0.200
ALCSM	0.800	0.800	0.200
ALL	0.800	0.800	0.400

\*\*\*\*\*

GAMMA PARAMETERS

GAMMAS ARE UNCONDITIONAL PROBABILITIES OF MEMBERSHIP IN EACH LATENT CLASS

OF THE STATIC LATENT VARIABLE

Male 0.500  
 Female 0.500

\*\*\*\*\*

DELTA PARAMETERS

DELTAS ARE PROBABILITIES OF LATENT STATUS MEMBERSHIP CONDITIONAL ON LATENT CLASS

DELTA PARAMETERS FOR LATENT CLASS "Male "

TIME 1  
 NOUSE 0.600  
 SMONLY 0.100  
 ALONLY 0.100  
 ALCSM 0.100  
 ALL 0.100

DELTA PARAMETERS FOR LATENT CLASS "Female "

TIME 1  
 NOUSE 0.600  
 SMONLY 0.100  
 ALONLY 0.100  
 ALCSM 0.100  
 ALL 0.100

\*\*\*\*\*

TAU PARAMETERS  
 TAUS ARE PROBABILITIES OF LATENT STATUS MEMBERSHIP AT TIME T+1  
 (COLUMNS)

CONDITIONAL ON LATENT STATUS MEMBERSHIP AT TIME T (ROWS)  
 AND ON LATENT CLASS MEMBERSHIP

(1X,A8,0005(F5.3,1x))

TRANSITION PROBABILITIES FOR LATENT CLASS "Male "

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 1  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.500	0.100	0.200	0.100	0.100
SMONLY	0.000	0.400	0.200	0.200	0.200
ALONLY	0.000	0.000	0.500	0.200	0.300
ALCSM	0.000	0.000	0.000	0.500	0.500
ALL	0.000	0.000	0.000	0.000	1.000

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 3

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.500	0.100	0.200	0.100	0.100
SMONLY	0.000	0.400	0.200	0.200	0.200
ALONLY	0.000	0.000	0.500	0.200	0.300
ALCSM	0.000	0.000	0.000	0.500	0.500
ALL	0.000	0.000	0.000	0.000	1.000

TRANSITION PROBABILITIES FOR LATENT CLASS "Female "

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 1  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.500	0.100	0.200	0.100	0.100
SMONLY	0.000	0.400	0.200	0.200	0.200
ALONLY	0.000	0.000	0.500	0.200	0.300
ALCSM	0.000	0.000	0.000	0.500	0.500
ALL	0.000	0.000	0.000	0.000	1.000

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 3

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.500	0.100	0.200	0.100	0.100
SMONLY	0.000	0.400	0.200	0.200	0.200
ALONLY	0.000	0.000	0.500	0.200	0.300
ALCSM	0.000	0.000	0.000	0.500	0.500
ALL	0.000	0.000	0.000	0.000	1.000

\*\*\*\*\*

ITERATION HISTORY

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STARTING G-SQUARED= 1450.303

ITER- ATION	MAD	ITER- ATION	MAD	ITER- ATION	MAD
1	.0292846925	2	.0135360129	3	.0099296022
4	.0080879475	5	.0069273243	6	.0062962571
7	.0057901429	8	.0054258402	9	.0051117213
10	.0048087516	11	.0045238549	12	.0042603550
13	.0040191525	14	.0037968760	15	.0035920644
16	.0034068270	17	.0032423023	18	.0030898270
19	.0029484705	20	.0028157520	21	.0026909254
22	.0025733351	23	.0024624081	24	.0023589934
25	.0022611923	26	.0021685976	27	.0020808899

<several pages of iteration history omitted>

3790	.0000010226	3791	.0000010219	3792	.0000010211
3793	.0000010204	3794	.0000010196	3795	.0000010189
3796	.0000010182	3797	.0000010174	3798	.0000010167
3799	.0000010159	3800	.0000010152	3801	.0000010145
3802	.0000010137	3803	.0000010130	3804	.0000010122
3805	.0000010115	3806	.0000010108	3807	.0000010100
3808	.0000010093	3809	.0000010086	3810	.0000010078
3811	.0000010071	3812	.0000010064	3813	.0000010056
3814	.0000010049	3815	.0000010042	3816	.0000010034
3817	.0000010027	3818	.0000010020	3819	.0000010013
3820	.0000010005	3821	.0000009998		

\*\*\*\*\*

G-SQUARED= 863.527 WITH 991 DEGREES OF FREEDOM

**\*\*WARNING\*\***: BE SURE TO INTERPRET THE LATENT CLASSES CAREFULLY  
 BASED ON THE ESTIMATED RHO PARAMETERS REPORTED BELOW.  
 YOU MAY WISH TO CHANGE THE LABELS YOU PREVIOUSLY  
 ASSIGNED TO THE LATENT CLASSES IN ORDER TO MAKE THEM  
 CONSISTENT WITH YOUR INTERPRETATION.

\*\*\*\*\*

LITTLE RHO PARAMETERS  
 LITTLE RHOS ARE PROBABILITIES OF RESPONSES  
 TO ITEMS MEASURING THE STATIC LATENT VARIABLE  
 CONDITIONAL ON LATENT CLASS MEMBERSHIP

RESPONSE CATEGORY 1

	G M
	E A
	N L
	D E
	E
	R
Male	1.000
Female	0.000

RESPONSE CATEGORY 2

	G F
	E E
	N M
	D A
	E L
	R E
Male	0.000
Female	1.000

**\*\*WARNING\*\***: BE SURE TO INTERPRET THE LATENT STATUSES CAREFULLY  
 BASED ON THE ESTIMATED RHO PARAMETERS REPORTED BELOW.  
 YOU MAY WISH TO CHANGE THE LABELS YOU PREVIOUSLY

ASSIGNED

TO THE LATENT STATUSES IN ORDER TO MAKE THEM  
 CONSISTENT WITH YOUR INTERPRETATION.

\*\*\*\*\*

BIG RHO PARAMETERS  
 BIG RHOS ARE PROBABILITIES OF RESPONSES  
 TO ITEMS MEASURING THE DYNAMIC LATENT VARIABLE  
 CONDITIONAL ON LATENT STATUS, LATENT CLASS, AND TIME

RHO PARAMETERS FOR LATENT CLASS "Male " AT TIME 1

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.897	0.849	0.981
SMONLY	0.246	0.849	0.981
ALONLY	0.897	0.121	0.981
ALCSM	0.246	0.121	0.718
ALL	0.246	0.121	0.543

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.103	0.151	0.019
SMONLY	0.754	0.151	0.019
ALONLY	0.103	0.879	0.019
ALCSM	0.754	0.879	0.282
ALL	0.754	0.879	0.457

RHO PARAMETERS FOR LATENT CLASS "Male " AT TIME 2

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.897	0.849	0.981
SMONLY	0.246	0.849	0.981
ALONLY	0.897	0.121	0.981
ALCSM	0.246	0.121	0.718
ALL	0.246	0.121	0.543

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.103	0.151	0.019
SMONLY	0.754	0.151	0.019
ALONLY	0.103	0.879	0.019
ALCSM	0.754	0.879	0.282
ALL	0.754	0.879	0.457

RHO PARAMETERS FOR LATENT CLASS "Male " AT TIME 3

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.897	0.849	0.981
SMONLY	0.246	0.849	0.981
ALONLY	0.897	0.121	0.981
ALCSM	0.246	0.121	0.718
ALL	0.246	0.121	0.543

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.103	0.151	0.019
SMONLY	0.754	0.151	0.019
ALONLY	0.103	0.879	0.019
ALCSM	0.754	0.879	0.282
ALL	0.754	0.879	0.457

RHO PARAMETERS FOR LATENT CLASS "Female " AT TIME 1

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.897	0.849	0.981
SMONLY	0.246	0.849	0.981
ALONLY	0.897	0.121	0.981
ALCSM	0.246	0.121	0.718
ALL	0.246	0.121	0.543

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.103	0.151	0.019
SMONLY	0.754	0.151	0.019
ALONLY	0.103	0.879	0.019
ALCSM	0.754	0.879	0.282
ALL	0.754	0.879	0.457



RHO PARAMETERS FOR LATENT CLASS "Female " AT TIME 2

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.897	0.849	0.981
SMONLY	0.246	0.849	0.981
ALONLY	0.897	0.121	0.981
ALCSM	0.246	0.121	0.718
ALL	0.246	0.121	0.543

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.103	0.151	0.019
SMONLY	0.754	0.151	0.019
ALONLY	0.103	0.879	0.019
ALCSM	0.754	0.879	0.282
ALL	0.754	0.879	0.457

RHO PARAMETERS FOR LATENT CLASS "Female " AT TIME 3

RESPONSE CATEGORY 1

	S N	D N	M N
	M O	R O	A O
	O	I	R
	K	N	I
	E	K	J
NOUSE	0.897	0.849	0.981
SMONLY	0.246	0.849	0.981
ALONLY	0.897	0.121	0.981
ALCSM	0.246	0.121	0.718
ALL	0.246	0.121	0.543

RESPONSE CATEGORY 2

	S Y	D Y	M Y
	M E	R E	A E
	O S	I S	R S
	K	N	I
	E	K	J
NOUSE	0.103	0.151	0.019
SMONLY	0.754	0.151	0.019
ALONLY	0.103	0.879	0.019
ALCSM	0.754	0.879	0.282
ALL	0.754	0.879	0.457

\*\*\*\*\*

GAMMA PARAMETERS

GAMMAS ARE UNCONDITIONAL PROBABILITIES OF MEMBERSHIP IN EACH LATENT CLASS

OF THE STATIC LATENT VARIABLE

Male	0.501
Female	0.499

\*\*\*\*\*

DELTA PARAMETERS

DELTAS ARE PROBABILITIES OF LATENT STATUS MEMBERSHIP CONDITIONAL ON LATENT CLASS

DELTA PARAMETERS FOR LATENT CLASS "Male "

	TIME 1	TIME 2	TIME 3
NOUSE	0.760	0.565	0.368
SMONLY	0.014	0.027	0.009
ALONLY	0.017	0.036	0.088
ALCSM	0.107	0.266	0.282
ALL	0.101	0.106	0.253

DELTA PARAMETERS FOR LATENT CLASS "Female "

	TIME 1	TIME 2	TIME 3
NOUSE	0.760	0.565	0.368
SMONLY	0.014	0.027	0.009
ALONLY	0.017	0.036	0.088
ALCSM	0.107	0.266	0.282
ALL	0.101	0.106	0.253

\*\*\*\*\*

TAU PARAMETERS

TAUS ARE PROBABILITIES OF LATENT STATUS MEMBERSHIP AT TIME T+1  
(COLUMNS)

CONDITIONAL ON LATENT STATUS MEMBERSHIP AT TIME T (ROWS)  
AND ON LATENT CLASS MEMBERSHIP

(1X,A8,0005(F5.3,1x))

TRANSITION PROBABILITIES FOR LATENT CLASS "Male "

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 1  
COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.744	0.016	0.027	0.209	0.005
SMONLY	0.000	1.000	0.000	0.000	0.000
ALCONLY	0.000	0.000	0.911	0.000	0.089
ALCSM	0.000	0.000	0.000	1.000	0.000
ALL	0.000	0.000	0.000	0.000	1.000

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2  
COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 3

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.651	0.014	0.100	0.000	0.235
SMONLY	0.000	0.023	0.104	0.874	0.000
ALCONLY	0.000	0.000	0.808	0.000	0.192
ALCSM	0.000	0.000	0.000	0.971	0.029
ALL	0.000	0.000	0.000	0.000	1.000

(1X,A8,0005(F5.3,1x))

TRANSITION PROBABILITIES FOR LATENT CLASS "Female "

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 1  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.744	0.016	0.027	0.209	0.005
SMONLY	0.000	1.000	0.000	0.000	0.000
ALCONLY	0.000	0.000	0.911	0.000	0.089
ALCSM	0.000	0.000	0.000	1.000	0.000
ALL	0.000	0.000	0.000	0.000	1.000

ROWS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 2  
 COLUMNS REPRESENT LATENT STATUS MEMBERSHIP AT TIME 3

	N	S	A	A	A
	O	M	L	L	L
	U	O	C	C	L
	S	N	O	S	
	E	L	N	M	
		Y	L		
			Y		
NOUSE	0.651	0.014	0.100	0.000	0.235
SMONLY	0.000	0.023	0.104	0.874	0.000
ALCONLY	0.000	0.000	0.808	0.000	0.192
ALCSM	0.000	0.000	0.000	0.971	0.029
ALL	0.000	0.000	0.000	0.000	1.000

\*\*\*\*\*

EXPECTED CELL FREQUENCIES AND RESIDUALS

	OBS	EXP	RESID	PEARSON
1111111111	25	30.1903	-5.1903	-0.9446
2111111111	44	30.0543	13.9457	2.5438 *
1211111111	2	3.4932	-1.4932	-0.7989
2211111111	2	3.4774	-1.4774	-0.7923
1121111111	8	5.3743	2.6257	1.1326
2121111111	2	5.3501	-3.3501	-1.4484
1221111111	1	0.6276	0.3724	0.4701
2221111111	1	0.6248	0.3752	0.4747
1212111111	1	0.0682	0.9318	3.5690 *
1222111111	4	0.0153	3.9847	32.2398 *
1111211111	1	3.5601	-2.5601	-1.3568
2111211111	5	3.5440	1.4560	0.7734
1211211111	1	0.4486	0.5514	0.8234
2211211111	2	0.4465	1.5535	2.3247 *
1222211111	1	0.0114	0.9886	9.2413 *
1111121111	1	5.6400	-4.6400	-1.9538
2111121111	3	5.6146	-2.6146	-1.1034

1211121111	2	0.6582	1.3418	1.6538
2211121111	2	0.6553	1.3447	1.6612
1121121111	3	1.1685	1.8315	1.6943
2121121111	1	1.1632	-0.1632	-0.1514
2221121111	1	0.1773	0.8227	1.9536
1122121111	1	0.0300	0.9700	5.5998 *
1222121111	1	0.0260	0.9740	6.0390 *
1111221111	1	0.8856	0.1144	0.1216
2111221111	2	0.8816	1.1184	1.1911
2221221111	1	0.1579	0.8421	2.1189 *
1222221111	1	0.0724	0.9276	3.4486 *
1111112111	1	0.5895	0.4105	0.5346
2111112111	1	0.5868	0.4132	0.5393
1122112111	1	0.0027	0.9973	19.3110 *
2111122111	1	0.1382	0.8618	2.3185 *
1111222111	1	0.1075	0.8925	2.7214 *
2111222111	3	0.1071	2.8929	8.8414 *
1211222111	1	0.0221	0.9779	6.5852 *
2121222111	1	0.0431	0.9569	4.6073 *
1221222111	2	0.0743	1.9257	7.0668 *
1111111211	8	4.7395	3.2605	1.4977
2111111211	3	4.7181	-1.7181	-0.7910
2211111211	2	0.5709	1.4291	1.8913
1221111211	1	0.1207	0.8793	2.5316 *
1222111211	1	0.0119	0.9881	9.0668 *
1111211211	1	0.7442	0.2558	0.2965
2111211211	1	0.7409	0.2591	0.3010
2111121211	1	1.1117	-0.1117	-0.1059
2211121211	1	0.1506	0.8494	2.1885 *
1121121211	1	0.2765	0.7235	1.3758
1221121211	1	0.1649	0.8351	2.0568 *
1111221211	1	0.8434	0.1566	0.1706
2111221211	1	0.8396	0.1604	0.1751
1222221211	1	0.2213	0.7787	1.6551
1111111121	18	11.8366	6.1634	1.7915
2111111121	11	11.7833	-0.7833	-0.2282
2211111121	1	1.4372	-0.4372	-0.3647
1121111121	4	2.2668	1.7332	1.1512
2121111121	1	2.2566	-1.2566	-0.8365
1112111121	1	0.2305	0.7695	1.6025
1222111121	1	0.0287	0.9713	5.7322 *
1111211121	1	1.8948	-0.8948	-0.6500
2111211121	2	1.8863	0.1137	0.0828
2211211121	1	0.4591	0.5409	0.7983
1121211121	1	0.3985	0.6015	0.9528
2121211121	2	0.3967	1.6033	2.5454 *
1111121121	5	4.1969	0.8031	0.3920
2111121121	5	4.1780	0.8220	0.4021
2211121121	1	0.5375	0.4625	0.6309
1121121121	1	1.9703	-0.9703	-0.6913
2121121121	3	1.9615	1.0385	0.7415
2221121121	1	0.5397	0.4603	0.6265
1122121121	1	0.0916	0.9084	3.0022 *
2222121121	1	0.1736	0.8264	1.9836
1111221121	3	2.1851	0.8149	0.5513
2111221121	1	2.1752	-1.1752	-0.7968
1121221121	2	0.8426	1.1574	1.2608

1221221121	1	1.0645	-0.0645	-0.0626
2221221121	1	1.0597	-0.0597	-0.0580
1111212121	1	0.1271	0.8729	2.4482 *
1111122121	1	0.2951	0.7049	1.2975
1111222121	2	0.6988	1.3012	1.5565
2111222121	1	0.6957	0.3043	0.3649
1121222121	2	0.3001	1.6999	3.1029 *
2222222121	1	0.3137	0.6863	1.2252
1111111221	5	6.7611	-1.7611	-0.6773
2111111221	11	6.7306	4.2694	1.6456
2211111221	1	0.9379	0.0621	0.0642
1121111221	2	1.2770	0.7230	0.6399
2121111221	1	1.2712	-0.2712	-0.2405
1221111221	1	0.3050	0.6950	1.2584
1111211221	3	2.0751	0.9249	0.6421
2111211221	4	2.0657	1.9343	1.3458
2211211221	1	0.7352	0.2648	0.3088
2221211221	1	0.5409	0.4591	0.6242
1212211221	1	0.0436	0.9564	4.5795 *
1111121221	5	3.2079	1.7921	1.0006
2111121221	4	3.1935	0.8065	0.4513
1211121221	1	0.5277	0.4723	0.6503
1121121221	1	1.1394	-0.1394	-0.1306
2121121221	1	1.1343	-0.1343	-0.1261
1122121221	1	0.1860	0.8140	1.8878
1222121221	1	0.5245	0.4755	0.6566
1111221221	5	5.5356	-0.5356	-0.2277
2111221221	5	5.5107	-0.5107	-0.2176
1211221221	1	1.1229	-0.1229	-0.1159
2211221221	4	1.1178	2.8822	2.7261 *
1121221221	2	2.0112	-0.0112	-0.0079
2121221221	4	2.0021	1.9979	1.4120
1221221221	4	3.1952	0.8048	0.4502
2221221221	1	3.1808	-2.1808	-1.2228
1222221221	2	1.6075	0.3925	0.3096
1111222221	4	2.1235	1.8765	1.2877
1222222221	1	0.9679	0.0321	0.0326
2222222221	1	0.9635	0.0365	0.0372
2111111112	1	0.7763	0.2237	0.2539
1111211112	1	0.1135	0.8865	2.6314 *
1222211112	1	0.0061	0.9939	12.7334 *
2111221112	1	0.1156	0.8844	2.6005 *
1111111122	1	1.6414	-0.6414	-0.5006
2111111122	2	1.6340	0.3660	0.2864
2111121122	2	0.5873	1.4127	1.8433
2121121122	1	0.2147	0.7853	1.6947
1111121222	1	0.0443	0.9557	4.5418 *
1111111222	5	4.4671	0.5329	0.2521
2111111222	5	4.4470	0.5530	0.2622
1121111222	1	0.8302	0.1698	0.1863
2121111222	1	0.8265	0.1735	0.1909
2221111222	1	0.1749	0.8251	1.9728
1222111222	1	0.0457	0.9543	4.4629 *
2222111222	1	0.0455	0.9545	4.4739 *
2111211222	2	1.0274	0.9726	0.9596
2221211222	1	0.2743	0.7257	1.3858
1111121222	1	1.6243	-0.6243	-0.4898

2111121222	2	1.6170	0.3830	0.3012
2211121222	1	0.2660	0.7340	1.4229
1121121222	1	0.5672	0.4328	0.5747
2121121222	1	0.5646	0.4354	0.5794
1122121222	1	0.1111	0.8889	2.6672 *
1111221222	3	2.3158	0.6842	0.4496
2111221222	4	2.3054	1.6946	1.1161
2121221222	3	0.9586	2.0414	2.0851 *
1221221222	1	1.6819	-0.6819	-0.5258
2221221222	5	1.6743	3.3257	2.5702 *
1212221222	1	0.1399	0.8601	2.2996 *
2222221222	1	0.9722	0.0278	0.0282
1111222222	2	0.8979	1.1021	1.1631
2211222222	1	0.2323	0.7677	1.5928
1121222222	2	0.4893	1.5107	2.1597 *
1221222222	2	0.9954	1.0046	1.0069
2221222222	1	0.9909	0.0091	0.0091
2122222222	1	0.2191	0.7809	1.6680
1222222222	1	0.6663	0.3337	0.4089
2222222222	1	0.6633	0.3367	0.4135

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PROGRAM FINISHED: Sun Jun 13 00:58:16 1999  
Time: 12197097