

Appendix0_10

```
# NDNS analysis, data management -----

# Change the data path accordingly -----
setwd("/home/wangcc-me/Downloads/UKDA-6533-stata11_se/stata11_se/") # on Ubuntu
library(epiDisplay)
library(plyr)
library(tidyverse)

# Read the data into memory -----
library(haven)
data <- read_dta("ndns_rp_yr1-4a_foodleveldietarydata_uk.dta")
data56 <- read_dta("ndns_rp_yr5-6a_foodleveldietarydata.dta")
data78 <- read_dta("ndns_rp_yr7-8a_foodleveldietarydata.dta")
names(data)
names(data56)
names(data78)
names(data)[names(data) == "seriali"] <- "id"
names(data56)[names(data56) == "seriali"] <- "id"
names(data78)[names(data78) == "seriali"] <- "id"

# Extract the data we needed -----
df14d <- data[, c(113, 1, 2, 3, 5, 6, 7, 8, 9, 21, 24, 55, 57, 58,
  59, 60, 61, 62, 63, 64)]
var <- names(df14d)
df56d <- data56 %>% select(var)
df78d <- data78 %>% select(var)
dfs1 <- rbind(df14d, df56d, df78d)
dfs2 <- dfs1[dfs1$Age >= 19, ]
rm(data, data56, data78)
dfs2

# Calculate the time (minute and hour) when they eat -----

dfs2$MealTime_chr <- as.character(dfs2$MealTime)
dfs2$MealTime_hm <- unlist(strsplit(dfs2$MealTime_chr, " "))[c(FALSE,
  TRUE)]
dfs2$MealHourN <- as.numeric(unlist(strsplit(dfs2$MealTime_hm, ":"))[c(TRUE,
  FALSE, FALSE)])
dfs2$MealMinN <- as.numeric(unlist(strsplit(dfs2$MealTime_hm, ":"))[c(FALSE,
  TRUE, FALSE)])
dfs2$MealMinN0 <- (60 * dfs2$MealHourN) + dfs2$MealMinN
dfs3 <- dfs2[order(dfs2$id, dfs2$DayNo, dfs2$MealMinN0), ]
length(unique(dfs3$id)) ## number of participants = 6155

# Create a subset data with only the first observation of each
# participant -----
NDNS <- dfs3[!duplicated(dfs3$id), ]
with(NDNS, tab1(SurveyYear, graph = FALSE, decimal = 2))

# #SurveyYear :
```

	#	Frequency	Percent	Cum. percent
# NDNS Year 1	801	13.01	13.01	
# NDNS Year 2	812	13.19	26.21	
# NDNS Year 3	782	12.71	38.91	
# NDNS Year 4	1055	17.14	56.05	
# NDNS Year 5	625	10.15	66.21	
# NDNS Year 6	663	10.77	76.98	
# NDNS Year 7	703	11.42	88.40	
# NDNS Year 8	714	11.60	100.00	
# Total	6155	100.00	100.00	

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# create a variable combine id and day No -----
dfs3 <- dfs3 %>%
mutate(id_dy = paste(id, DayNo, sep = "D"))

# For each subject, the total energy/carbohydrate intake for each eating
# time can be calculated -----
old <- Sys.time()
Energy <- ddply(dfs3, .(id_dy, id, SurveyYear, DayNo, Age, Sex,
                      DiaryDaysCompleted, MealHourN, DayofWeek),
               summarise,
               Tot_Energ = sum(EnergykJ),
               Tot_Carb = sum(Carbohydrateg),
               Tot_Sugar = sum(Totalsugarsg),
               Tot_Starch = sum(Starchg))
new <- Sys.time() - old
print(new)
# Time difference of 3.876385 mins

rm(df14d, df56d, df78d, dfs2)

# Calculate the energy from total carbohydrates -----
Energy <- Energy %>%
  mutate(KJcarbo = Tot_Carb * 16) %>%
  mutate(CarKJpercentage = KJcarbo/Tot_Energ) %>%
  mutate(Carbo = cut(CarKJpercentage, breaks = c(0, 0.26, 0.75, 2),
                    right = FALSE)) %>% mutate(Carbo2 = cut(CarKJpercentage, breaks = c(0,
                    0.26, 2), right = FALSE))
Energy0 <- Energy[!(Energy$Tot_Energ == 0), ]
# some food consumption does not contain any carbohydrates
Energy0$Carbo <- factor(Energy0$Carbo, labels = c("Low_carb", "Med_carb",
"High_carb"))
Energy0$Carbo2 <- factor(Energy0$Carbo2, labels = c("Low_carb", "Med_or_high_carb"))

# Generate data sets for each day -----
dta_day1 <- Energy0 %>%
  filter(DayNo == 1) %>%
  select(c("id", "Age",
           "Sex", "DayofWeek", "MealHourN", "Carbo", "Carbo2")) %>%
  mutate(DayofWeek = factor(DayofWeek,
                           levels = c("Monday", "Tuesday", "Wednesday", "Thursday", "Friday",
                           "Saturday", "Sunday")))

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dta_day2 <- Energy0 %>%
  filter(DayNo == 2) %>%
  select(c("id", "Age",
           "Sex", "DayofWeek", "MealHourN", "Carbo", "Carbo2")) %>%
  mutate(DayofWeek = factor(DayofWeek,
                             levels = c("Monday", "Tuesday", "Wednesday", "Thursday", "Friday",
                                           "Saturday", "Sunday")))

dta_day3 <- Energy0 %>%
  filter(DayNo == 3) %>%
  select(c("id", "Age",
           "Sex", "DayofWeek", "MealHourN", "Carbo", "Carbo2")) %>%
  mutate(DayofWeek = factor(DayofWeek,
                             levels = c("Monday", "Tuesday", "Wednesday", "Thursday", "Friday",
                                           "Saturday", "Sunday")))

dta_day4 <- Energy0 %>%
  filter(DayNo == 4) %>%
  select(c("id", "Age",
           "Sex", "DayofWeek", "MealHourN", "Carbo", "Carbo2")) %>%
  mutate(DayofWeek = factor(DayofWeek,
                             levels = c("Monday", "Tuesday", "Wednesday", "Thursday", "Friday",
                                           "Saturday", "Sunday")))

vecid1 <- unique(dta_day1$id) # n = 6153
vecid2 <- unique(dta_day2$id) # n = 6153
vecid3 <- unique(dta_day3$id) # n = 6151
vecid4 <- unique(dta_day4$id) # n = 6026

Noday1 <- setdiff(vecid, vecid1) # two subjects did not have day 1 data
Noday2 <- setdiff(vecid, vecid2) # two subjects did not have day 2 data
Noday3 <- setdiff(vecid, vecid3) # four subjects did not have day 3 data
Noday4 <- setdiff(vecid, vecid4) # 129 subjects did not have day 4 data

# Transform the data shape from long to wide -----
dta_d1_wide <- dta_day1[, -7] %>%
  spread(key = MealHourN, value = Carbo)
names(dta_d1_wide)[5:28] <- paste(rep("H", 24), 0:23, sep = "")

dta_d2_wide <- dta_day2[, -7] %>%
  spread(key = MealHourN, value = Carbo)
names(dta_d2_wide)[5:28] <- paste(rep("H", 24), 0:23, sep = "")

dta_d3_wide <- dta_day3[, -7] %>%
  spread(key = MealHourN, value = Carbo)
names(dta_d3_wide)[5:28] <- paste(rep("H", 24), 0:23, sep = "")

dta_d4_wide <- dta_day4[, -7] %>%
  spread(key = MealHourN, value = Carbo)
names(dta_d4_wide)[5:28] <- paste(rep("H", 24), 0:23, sep = "")

# recode NA to not eating -----
for (i in 5:ncol(dta_d1_wide))
  if (is.factor(dta_d1_wide[, i])) levels(dta_d1_wide[,

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i]) <- c(levels(dta_d1_wide[, i]), "Not_eating")

dta_d1_wide[is.na(dta_d1_wide)] <- "Not_eating"

for (i in 5:ncol(dta_d2_wide))
  if (is.factor(dta_d2_wide[, i])) levels(dta_d2_wide[,
    i]) <- c(levels(dta_d2_wide[, i]), "Not_eating")

dta_d2_wide[is.na(dta_d2_wide)] <- "Not_eating"

for (i in 5:ncol(dta_d3_wide))
  if (is.factor(dta_d3_wide[, i])) levels(dta_d3_wide[,
    i]) <- c(levels(dta_d3_wide[, i]), "Not_eating")

dta_d3_wide[is.na(dta_d3_wide)] <- "Not_eating"

for (i in 5:ncol(dta_d4_wide))
  if (is.factor(dta_d4_wide[, i])) levels(dta_d4_wide[,
    i]) <- c(levels(dta_d4_wide[, i]), "Not_eating")

dta_d4_wide[is.na(dta_d4_wide)] <- "Not_eating"

```

Mplus VERSION 7.4
MUTHEN & MUTHEN
07/28/2018 9:55 AM

INPUT INSTRUCTIONS

TITLE: 3-class at level 1 (CW), 3-classes at level 2 (CB) random effects model - non-pa
ordered polytomous variables for carb intake at each time slot over four
days of NDNS survey 2008/09 - 2015/16
variable 0 = not eating
1 = eating & carb provided < 50% calorie
2 = eating & carb provided >= 50% calorie

DATA: File is H:\summer_project\Mplus\TimeSlots\NDNS_Tslots.dat;

VARIABLE: NAMES = id id_dy Age Sex H6_9 H9_12 H12_14 H14_17 H17_20
H20_22 H22_6;

USEVAR = H6_9 H9_12 H12_14 H14_17 H17_20
H20_22 H22_6;

auxiliary = Age Sex;

CATEGORICAL = H6_9 H9_12 H12_14 H14_17 H17_20
H20_22 H22_6;

CLUSTER = id;

IDVARIABLE = id_dy;

BETWEEN = CB;

```

    WITHIN = H6_9 H9_12 H12_14 H14_17 H17_20
             H20_22 H22_6;

```

```

    CLASSES = CB(3) CW(3);

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    MISSING are .;

```

ANALYSIS:

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type = mixture twolevel;
starts = 50 25;
process = 8(starts);

```

MODEL:

```

%within%
%overall%
%between%
%overall%
CW ON CB;

```

Savedata:

```

file is H:\summer_project\Mplus\TimeSlots\Multilevel\NDNSslot_CW3CB3.txt;
save is cprob;
format is free;

```

3-class at level 1 (CW), 3-classes at level 2 (CB) random effects model - non-par
ordered polytomous variables for carb intake at each time slot over four
days of NDNS survey 2008/09 - 2015/16
variable 0 = not eating
1 = eating & carb provided < 50% calorie
2 = eating & carb provided >= 50% calorie

SUMMARY OF ANALYSIS

Number of groups	1
Number of observations	24483
Number of dependent variables	7
Number of independent variables	0
Number of continuous latent variables	0
Number of categorical latent variables	2

Observed dependent variables

Binary and ordered categorical (ordinal)

H6_9	H9_12	H12_14	H14_17	H17_20	H20_22
H22_6					

Observed auxiliary variables

AGE SEX

Categorical latent variables

CB CW

Variables with special functions

Cluster variable ID
ID variable ID_DY

Within variables

H6_9 H9_12 H12_14 H14_17 H17_20 H20_22
H22_6

Estimator MLR
Information matrix OBSERVED
Optimization Specifications for the Quasi-Newton Algorithm for
Continuous Outcomes
Maximum number of iterations 100
Convergence criterion 0.100D-05
Optimization Specifications for the EM Algorithm
Maximum number of iterations 500
Convergence criteria
Loglikelihood change 0.100D-02
Relative loglikelihood change 0.100D-05
Derivative 0.100D-02
Optimization Specifications for the M step of the EM Algorithm for
Categorical Latent variables
Number of M step iterations 1
M step convergence criterion 0.100D-02
Basis for M step termination ITERATION
Optimization Specifications for the M step of the EM Algorithm for
Censored, Binary or Ordered Categorical (Ordinal), Unordered
Categorical (Nominal) and Count Outcomes
Number of M step iterations 1
M step convergence criterion 0.100D-02
Basis for M step termination ITERATION
Maximum value for logit thresholds 15
Minimum value for logit thresholds -15
Minimum expected cell size for chi-square 0.100D-01
Maximum number of iterations for H1 2000
Convergence criterion for H1 0.100D-03
Optimization algorithm EMA
Integration Specifications
Type STANDARD
Number of integration points 15
Dimensions of numerical integration 0
Adaptive quadrature ON
Random Starts Specifications
Number of initial stage random starts 50
Number of final stage optimizations 25

Number of initial stage iterations	10
Initial stage convergence criterion	0.100D+01
Random starts scale	0.500D+01
Random seed for generating random starts	0
Parameterization	LOGIT
Link	LOGIT
Cholesky	OFF

Input data file(s)
H:\summer_project\Mplus\TimeSlots\NDNS_Tslots.dat
Input data format FREE

SUMMARY OF DATA

Number of missing data patterns	1
Number of y missing data patterns	0
Number of u missing data patterns	1
Number of clusters	6155

COVARIANCE COVERAGE OF DATA

Minimum covariance coverage value 0.100

UNIVARIATE PROPORTIONS AND COUNTS FOR CATEGORICAL VARIABLES

H6_9		
Category 1	0.313	7655.000
Category 2	0.184	4500.000
Category 3	0.504	12328.000
H9_12		
Category 1	0.222	5447.000
Category 2	0.295	7227.000
Category 3	0.482	11809.000
H12_14		
Category 1	0.195	4783.000
Category 2	0.454	11112.000
Category 3	0.351	8588.000
H14_17		
Category 1	0.283	6926.000
Category 2	0.338	8277.000
Category 3	0.379	9280.000
H17_20		
Category 1	0.124	3043.000
Category 2	0.582	14240.000
Category 3	0.294	7200.000
H20_22		
Category 1	0.356	8722.000
Category 2	0.363	8898.000
Category 3	0.280	6863.000
H22_6		

Category 1	0.666	16295.000
Category 2	0.169	4144.000
Category 3	0.165	4044.000

RANDOM STARTS RESULTS RANKED FROM THE BEST TO THE WORST LOGLIKELIHOOD VALUES

Final stage loglikelihood values at local maxima, seeds, and initial stage start numbers:

-166348.815	153942	31
-166348.815	573096	20
-166348.815	253358	2
-166348.816	318230	46
-166348.816	246261	38
-166348.873	285380	1
-166348.908	903420	5
-166349.394	120506	45
-166349.394	966014	37
-166349.394	207896	25
-166349.395	195873	6
-166349.513	68985	17
-166349.514	366706	29
-166352.737	76974	16
-166357.057	127215	9
-166482.723	533738	11
-166495.844	645664	39
-166668.918	372176	23

THE BEST LOGLIKELIHOOD VALUE HAS BEEN REPLICATED. RERUN WITH AT LEAST TWICE THE RANDOM STARTS TO CHECK THAT THE BEST LOGLIKELIHOOD IS STILL OBTAINED AND REPLICATED.

THE MODEL ESTIMATION TERMINATED NORMALLY

MODEL FIT INFORMATION

Number of Free Parameters	134
---------------------------	-----

Loglikelihood

H0 Value	-166348.815
H0 Scaling Correction Factor for MLR	1.8182

Information Criteria

Akaike (AIC)	332965.630
Bayesian (BIC)	334051.799
Sample-Size Adjusted BIC (n* = (n + 2) / 24)	333625.950

MODEL RESULTS USE THE LATENT CLASS VARIABLE ORDER

CB CW

Latent Class Variable Patterns

CB Class	CW Class
1	1
1	2
1	3
2	1
2	2
2	3
3	1
3	2
3	3

FINAL CLASS COUNTS AND PROPORTIONS FOR THE LATENT CLASS PATTERNS
BASED ON ESTIMATED POSTERIOR PROBABILITIES

Latent Class
Pattern

1	1	4050.97975	0.16546
1	2	1561.55249	0.06378
1	3	1286.46696	0.05255
2	1	2746.94031	0.11220
2	2	3011.00217	0.12298
2	3	1341.59686	0.05480
3	1	2748.25320	0.11225
3	2	4770.55950	0.19485
3	3	2965.64876	0.12113

FINAL CLASS COUNTS AND PROPORTIONS FOR EACH LATENT CLASS VARIABLE
BASED ON ESTIMATED POSTERIOR PROBABILITIES

Latent Class
Variable Class

CB	1	6898.99902	0.28179
	2	7099.53906	0.28998
	3	10484.46094	0.42823
CW	1	9546.17285	0.38991
	2	9343.11426	0.38162
	3	5593.71240	0.22847

FINAL CLASS COUNTS AND PROPORTIONS FOR THE LATENT CLASS PATTERNS

Class Counts and Proportions

FINAL CLASS COUNTS AND PROPORTIONS FOR EACH LATENT CLASS VARIABLE
BASED ON THEIR MOST LIKELY LATENT CLASS PATTERN

CLASSIFICATION QUALITY

Average Latent Class Probabilities for Most Likely Latent Class Pattern (Row)
by Latent Class Pattern (Column)

10

	1	2	3	4	5	6	7	8	9
1	0.720	0.091	0.073	0.016	0.032	0.004	0.005	0.033	0.025
2	0.183	0.609	0.098	0.005	0.002	0.030	0.040	0.005	0.027
3	0.211	0.084	0.629	0.008	0.005	0.007	0.011	0.036	0.009
4	0.019	0.004	0.002	0.692	0.184	0.051	0.011	0.034	0.003
5	0.042	0.001	0.001	0.158	0.709	0.045	0.001	0.035	0.009
6	0.012	0.037	0.013	0.065	0.084	0.702	0.042	0.003	0.042
7	0.011	0.029	0.004	0.012	0.002	0.022	0.641	0.126	0.153
8	0.026	0.003	0.009	0.025	0.024	0.001	0.115	0.675	0.123
9	0.046	0.024	0.004	0.003	0.010	0.018	0.079	0.174	0.642

MODEL RESULTS

	Estimate	S.E.	Est./S.E.	Two-Tailed P-Value
Within Level				
Latent Class Pattern 1 1				
Thresholds				
H6_9\$1	-0.718	0.218	-3.294	0.001
H6_9\$2	0.973	0.299	3.258	0.001
H9_12\$1	-2.516	0.463	-5.433	0.000
H9_12\$2	0.675	0.132	5.118	0.000
H12_14\$1	-1.025	0.145	-7.057	0.000
H12_14\$2	1.240	0.116	10.725	0.000
H14_17\$1	-1.566	0.149	-10.520	0.000
H14_17\$2	1.090	0.100	10.909	0.000
H17_20\$1	-1.998	0.125	-16.000	0.000
H17_20\$2	1.549	0.100	15.556	0.000
H20_22\$1	-0.933	0.085	-10.914	0.000
H20_22\$2	1.829	0.103	17.770	0.000
H22_6\$1	0.253	0.083	3.046	0.002
H22_6\$2	2.308	0.117	19.691	0.000

Latent Class Pattern 1 2

Thresholds				
H6_9\$1	-4.021	1.788	-2.249	0.025
H6_9\$2	-0.115	0.259	-0.445	0.656
H9_12\$1	0.167	0.373	0.448	0.654
H9_12\$2	2.142	0.586	3.657	0.000
H12_14\$1	-3.210	1.518	-2.115	0.034
H12_14\$2	0.858	0.167	5.124	0.000
H14_17\$1	0.044	0.384	0.114	0.909
H14_17\$2	1.617	0.293	5.509	0.000
H17_20\$1	-2.109	0.390	-5.409	0.000
H17_20\$2	1.399	0.196	7.126	0.000
H20_22\$1	-0.367	0.174	-2.109	0.035

H20_22\$2	2.347	0.382	6.151	0.000
H22_6\$1	0.754	0.259	2.912	0.004
H22_6\$2	2.542	0.264	9.646	0.000

Latent Class Pattern 1 3

Thresholds

H6_9\$1	-15.000	0.000	999.000	999.000
H6_9\$2	2.357	0.783	3.011	0.003
H9_12\$1	-1.433	0.372	-3.850	0.000
H9_12\$2	-0.604	0.279	-2.166	0.030
H12_14\$1	-1.988	0.257	-7.749	0.000
H12_14\$2	0.524	0.125	4.209	0.000
H14_17\$1	-1.027	0.232	-4.436	0.000
H14_17\$2	0.274	0.131	2.087	0.037
H17_20\$1	-2.665	0.310	-8.605	0.000
H17_20\$2	0.707	0.112	6.322	0.000
H20_22\$1	-0.527	0.152	-3.462	0.001
H20_22\$2	0.702	0.138	5.102	0.000
H22_6\$1	1.119	0.185	6.062	0.000
H22_6\$2	1.748	0.183	9.544	0.000

Latent Class Pattern 2 1

Thresholds

H6_9\$1	1.663	0.199	8.370	0.000
H6_9\$2	1.839	0.198	9.274	0.000
H9_12\$1	-2.150	0.281	-7.643	0.000
H9_12\$2	-0.869	0.140	-6.190	0.000
H12_14\$1	-1.978	0.191	-10.349	0.000
H12_14\$2	0.323	0.078	4.139	0.000
H14_17\$1	0.237	0.183	1.293	0.196
H14_17\$2	0.782	0.123	6.352	0.000
H17_20\$1	-2.936	0.428	-6.853	0.000
H17_20\$2	0.632	0.081	7.807	0.000
H20_22\$1	0.028	0.142	0.194	0.846
H20_22\$2	0.868	0.086	10.145	0.000
H22_6\$1	0.658	0.109	6.010	0.000
H22_6\$2	1.326	0.100	13.215	0.000

Latent Class Pattern 2 2

Thresholds

H6_9\$1	1.640	0.171	9.619	0.000
H6_9\$2	1.906	0.179	10.678	0.000
H9_12\$1	-1.954	0.347	-5.636	0.000
H9_12\$2	-0.360	0.127	-2.842	0.004
H12_14\$1	-0.016	0.189	-0.084	0.933
H12_14\$2	0.948	0.135	7.029	0.000
H14_17\$1	-1.906	0.301	-6.327	0.000
H14_17\$2	0.371	0.080	4.614	0.000
H17_20\$1	-0.812	0.116	-7.030	0.000
H17_20\$2	0.910	0.089	10.259	0.000
H20_22\$1	-0.742	0.089	-8.318	0.000

H20_22\$2	0.998	0.085	11.705	0.000
H22_6\$1	0.298	0.083	3.608	0.000
H22_6\$2	1.337	0.099	13.475	0.000

Latent Class Pattern 2 3

Thresholds

H6_9\$1	-1.072	0.500	-2.144	0.032
H6_9\$2	-0.309	0.346	-0.892	0.372
H9_12\$1	2.441	1.044	2.339	0.019
H9_12\$2	3.599	1.983	1.815	0.069
H12_14\$1	-1.029	0.211	-4.880	0.000
H12_14\$2	0.603	0.123	4.913	0.000
H14_17\$1	-0.010	0.243	-0.041	0.967
H14_17\$2	0.784	0.157	4.977	0.000
H17_20\$1	-0.953	0.203	-4.684	0.000
H17_20\$2	0.779	0.135	5.784	0.000
H20_22\$1	-0.105	0.210	-0.500	0.617
H20_22\$2	1.203	0.135	8.914	0.000
H22_6\$1	0.582	0.299	1.950	0.051
H22_6\$2	1.370	0.206	6.653	0.000

Latent Class Pattern 3 1

Thresholds

H6_9\$1	-4.593	1.699	-2.703	0.007
H6_9\$2	-2.975	0.428	-6.957	0.000
H9_12\$1	-0.322	0.207	-1.553	0.120
H9_12\$2	0.398	0.363	1.095	0.274
H12_14\$1	-5.060	3.668	-1.380	0.168
H12_14\$2	0.307	0.100	3.080	0.002
H14_17\$1	0.186	0.530	0.351	0.726
H14_17\$2	0.317	0.245	1.295	0.195
H17_20\$1	-4.019	0.957	-4.199	0.000
H17_20\$2	0.747	0.093	7.987	0.000
H20_22\$1	-0.233	0.132	-1.767	0.077
H20_22\$2	0.607	0.109	5.571	0.000
H22_6\$1	1.304	0.146	8.918	0.000
H22_6\$2	1.850	0.160	11.579	0.000

Latent Class Pattern 3 2

Thresholds

H6_9\$1	-1.232	0.195	-6.305	0.000
H6_9\$2	-0.858	0.169	-5.068	0.000
H9_12\$1	-4.377	1.937	-2.260	0.024
H9_12\$2	-1.488	0.316	-4.717	0.000
H12_14\$1	-1.727	0.227	-7.611	0.000
H12_14\$2	0.302	0.082	3.666	0.000
H14_17\$1	-1.834	0.237	-7.730	0.000
H14_17\$2	-0.294	0.186	-1.582	0.114
H17_20\$1	-2.588	0.487	-5.313	0.000
H17_20\$2	0.631	0.062	10.187	0.000
H20_22\$1	-0.920	0.078	-11.852	0.000

H20_22\$2	0.462	0.073	6.308	0.000
H22_6\$1	0.640	0.119	5.361	0.000
H22_6\$2	1.162	0.129	9.039	0.000

Latent Class Pattern 3 3

Thresholds

H6_9\$1	-4.941	5.813	-0.850	0.395
H6_9\$2	-2.680	0.887	-3.024	0.002
H9_12\$1	-0.765	0.640	-1.195	0.232
H9_12\$2	1.164	0.920	1.265	0.206
H12_14\$1	-1.415	0.439	-3.226	0.001
H12_14\$2	0.566	0.085	6.626	0.000
H14_17\$1	-2.052	0.650	-3.158	0.002
H14_17\$2	0.612	0.210	2.909	0.004
H17_20\$1	-1.627	0.427	-3.810	0.000
H17_20\$2	0.713	0.103	6.935	0.000
H20_22\$1	-0.850	0.329	-2.585	0.010
H20_22\$2	0.685	0.134	5.104	0.000
H22_6\$1	1.237	0.195	6.349	0.000
H22_6\$2	1.893	0.179	10.582	0.000

Between Level

Categorical Latent Variables

Within Level

Intercepts

CW#1	-0.076	0.366	-0.208	0.835
CW#2	0.475	0.309	1.539	0.124

Between Level

CW#1 ON

CB#1	1.223	0.473	2.585	0.010
CB#2	0.793	0.441	1.796	0.073

CW#2 ON

CB#1	-0.282	0.535	-0.526	0.599
CB#2	0.333	0.455	0.733	0.464

Means

CB#1	-0.417	0.100	-4.178	0.000
CB#2	-0.386	0.067	-5.770	0.000

QUALITY OF NUMERICAL RESULTS

Condition Number for the Information Matrix (ratio of smallest to largest eigenvalue)	0.428E-04
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SAVEDATA INFORMATION

Save file

H:\summer_project\Mplus\TimeSlots\Multilevel\NDNSslot_CW3CB3.txt

Order of variables

H6_9
H9_12
H12_14
H14_17
H17_20
H20_22
H22_6
ID_DY
AGE
SEX
CPROB1
CPROB2
CPROB3
CPROB4
CPROB5
CPROB6
CPROB7
CPROB8
CPROB9
CB
CW
MLCJOINT
ID

Save file format Free

Save file record length 10000

DIAGRAM INFORMATION

Mplus diagrams are currently not available for Mixture analysis.
No diagram output was produced.

Beginning Time: 09:55:10
Ending Time: 10:02:01
Elapsed Time: 00:06:51

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