

Modeling Relationship between Alcohol Policy  
Perception and Alcohol Consumption in '01  
Harvard College Alcohol Survey (CAS)

Youngsoo Baek, Michael Christensen, and Yufeng Jiang

# Objective

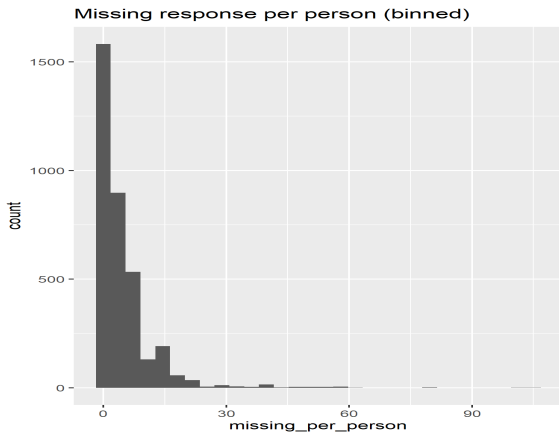
- ▶ Data: 2001 Harvard College Alcohol Study (CAS)
  - ▶ 10904 participants (unknown response rate)
- ▶ Investigate the correlation structure between **subjective beliefs about campus alcohol policy** and **objective measures of alcohol consumption**
- ▶ *Section B* for subjective questions, *Section C* for objective questions
- ▶ Standard survey modeling techniques: factor analysis, structural equations model, item response theory

# Data Processing

- ▶ Consistent ordering of responses
  - ▶ More stringent alcohol policy beliefs (1) -> Less stringent
  - ▶ Less alcohol consumption (1) -> More consumption
- ▶ Aggressive pruning of the variables before modeling
- ▶ Unreliable responses classified based on
  - ▶ Response to A7: A (alone) is not allowed with other responses (family/partner/roommate)
  - ▶ Response in Section C: participants who chose 1 in C10 and answered C11–C15, etc.
- ▶ WEIGHT01 used as sampling weights (intended for cross-sectional studies)

# Missing Responses

- ▶ Missing response rate adjusted for questions that only target certain demographic subgroups



## Missing Responses for Each Section

# Structural Equations Modeling (SEM)

- ▶ Survey responses  $X_i$  can be grouped together as repeated measurements of a lower-dimensional, latent *factors*: alcohol beliefs, alcohol consumption attitude, ...
- ▶ Factor analysis identifies the loadings  $\Lambda$  of latent variable  $\eta_i$ .

$$X_i = \Lambda \eta_i + \epsilon_i, \quad \eta_i \sim N(0, 1), \quad \epsilon_i \sim N(0, 1)$$

- ▶ Structural Equations Models extend factor analysis by specifying within-question correlations and regressing  $\eta_i$  on predictors.
  - ▶ All of our model predictors are directly observed rather than “manifested” by questions: age, gender, ...
- ▶ Causal interpretation is **not** necessary (though often made!).

# Graphical Representation of the Model

...PLOT...

# Modeling Challenges

- ▶ The model is clearly misspecified: Gaussian error assumption is made on ordered response
  - ▶ Asymptotic standard errors of factor loading estimators are valid for nonnormal factor analyses (Anderson and Amemiya, 1988)
  - ▶ In practice can cause lower goodness-of-fit
- ▶ Complete case analysis due to excessive computation in maximizing the full likelihood
- ▶ (ANYTHING ELSE??)



## Main Results

A MATRIX PLOT OF LOADINGS (FACTOR CORRELATION)  
WILL BE GREAT HERE, RATHER THAN ANY ESTIMATE  
FIGURES ...

# Model Diagnostics

- ▶ Various statistics to evaluate model fit in practice: TLI, BL89, CFI, RMSEA (Hu and Bentler, 1999)
- ▶ (INCLUDE RESULTS HERE ...)

# Interpretation

...

# Conclusion

- ▶ ...
- ▶ Limitations
  - ▶ Alternative approaches to account for ordered response
  - ▶ Theory-driven priors may improve fit of more complex models
  - ▶ Need information to correct for estimate biases

# Reference

- ▶ “Asymptotic Chi-Square Tests for a Large Class of Factor Analysis Models,” Anderson, T. W. and Amemiya, Y. *The Annals of Statistics*, 16(2), 1988.
- ▶ “Cutoff criteria for fit indexes in covariance structure analysis: Conventional criteria versus new alternatives,” Hu, L.-T. and Bentler, P. M. *Structural Equation Modeling: A Multidisciplinary Journal*, 6(1), 1999.