

# A Two-part Mixed-effects Model for Longitudinal Outcomes with "Don't Know" Category: Implication for Program Evaluation

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## Background

- "Don't Know (DK)" category has been increasingly used in surveys as an alternative for respondents
- Higher frequencies of DKs have been consistently reported among respondents with **vulnerable characteristics** in health-related research, especially socio-demographics
- **Challenges:** Although reducing item non-responses and guess work in some extent, DKs create unique challenges in data analysis and program evaluation, especially for longitudinal data
- **Current Limitations:** Common approaches such as treating DKs as missing data will not only lose meaningful information, but also lead to biases and loss of precision on the estimates of intervention effects



### Possible Reasons of DK

- Lack of knowledge
- Lack of involvement
- Lack of interest
- Satisficing
- Self-image protection
- ...

## Objectives

- Develop a **statistical model** for longitudinal responses with DK category
- Conduct **simulation studies** to illustrate the merits of the proposed model over other alternative methods
- Apply the proposed model in **program evaluation** and compare it with other methods, e.g., treating DKs as missing data

## Methods

### Model Development

- A **two-part mixed-effects model** will be developed (i: subject, t: time point):
  - One part characterizing DKs ( $\Pr(y_{it} = 'DK')$ ) over time using a logistic mixed model (LMM)
  - The other part modeling the evolving patterns in longitudinal outcomes ( $y_{it} \neq 'DK'$ ) using a non-linear mixed model (NLMM)
  - The random effects of two parts ( $\alpha_i$  &  $\beta_i$ ) are assumed to be correlated (correlation coefficient:  $\rho$ ) to account for associations between DKs and longitudinal outcomes

$$\begin{aligned} \text{LMM: } \text{logit}(\Pr(y_{it} = 'DK')) &= X\lambda + \alpha_i \\ \text{NLMM: } E(\eta(y_{it}) | y_{it} \neq 'DK') &= X\gamma + \beta_i \end{aligned} \quad \text{Correlation: } \begin{pmatrix} \alpha_i \\ \beta_i \end{pmatrix} \sim N\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma_\alpha^2 & \rho\sigma_\alpha\sigma_\beta \\ \rho\sigma_\alpha\sigma_\beta & \sigma_\beta^2 \end{pmatrix}\right)$$

### Simulation Studies

- Simulation studies will be conducted to illustrate the advantages of the proposed model on parameter estimates over other methods under **different conditions of sample size, DK proportions, and correlation strengths.**

### Application in Intervention Program

- The proposed model will be applied to evaluate an intervention program, and compared with other approaches treating DKs as missing data, including deleting DKs and multiple imputation of DKs
- Data: PROJECT 11, a school-based mental health prevention program in Manitoba.

## Significance

- Deeper understanding of **special item non-responses** like DKs in longitudinal research
- Improvements on the **conclusion quality** in program evaluation, and potential to identify **marginal / vulnerable populations** and inform more targeted preventions and interventions

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