# A multivariate skew-normal finite mixture model for analysis of infant development trajectories

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Summary: In studies of infant motor development, a crucial research goal is to identify latent classes of infants that experience delayed development, as this is a known risk factor for adverse outcomes later in life. However, there are a number of statistical challenges in modeling infant development: the data are typically skewed, exhibit intermittent missingness, and are highly correlated across the repeated measurements collected during infancy. Using data from the Nurture study, a cohort of over 600 mother-infant pairs followed from pregnancy to 12 months postpartum, we develop a flexible Bayesian latent class model for the analysis infant motor development. Our model has a number of attractive features. First, we adopt the multivariate skew normal distribution with class-specific parameters that accommodate the inherent correlation and skewness in the data. Second, we model the class membership probabilities using a novel Plya-Gamma data-augmentation scheme, thereby improving predictions of the class membership allocations. Lastly, we impute missing responses under missing at random assumption by drawing from appropriate conditional skew normal distributions. Bayesian inference is achieved through straightforward Gibbs sampling, and can be carried out in available software such as R. Through simulation studies, we show that the proposed model yields improved

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inferences over models that ignore skewness. In addition, our imputation method yields improvements compared to conventional missing data methods, including multiple imputation and complete or available case analysis. When applied to Nurture data, we identified two distinct development classes: one characterized by delayed U-shaped development and a higher percentage of male infants and another characterized by more steady development and a lower percentage of males. The classes also differed in terms of key demographic variables, such as infant race and maternal pre-pregnancy body mass index. These findings can aid investigators in targeting interventions during this critical early-life developmental window.

KEY WORDS: A key word; But another key word; Still another key word; Yet another key word.

- 1. Introduction
- 2. Model
- 2.1 Multivariate Skew Normal Regression
- 2.2 Latent Class Finite Mixture
- 2.3 Multinomial Regression on Class Probabilities
- 2.4 Conditional MVSN Imputation
- 3. Bayesian Inference
- 4. Simulation Studies
- 5. Application
- 6. Discussion

Put your final comments here.

## ACKNOWLEDGEMENTS

## SUPPLEMENTARY MATERIALS

### References

- Cox, D. R. (1972). Regression models and life tables (with discussion). *Journal of the Royal Statistical Society, Series B* **34**, 187–200.
- Hastie, T., Tibshirani, R., and Friedman, J. (2001). The Elements of Statistical Learning:

  Data Mining, Inference, and Prediction. New York: Springer.

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## APPENDIX

Full Conditional Distributions