

How to estimate a population proportion if data
are possibly subject to misclassification error?
The case of estimating contraceptive prevalence
based on self-reported usage.

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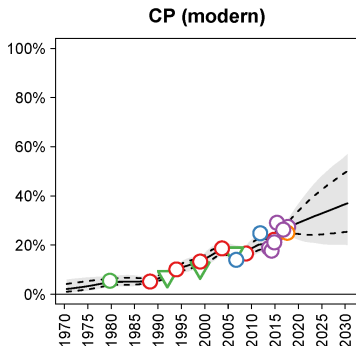
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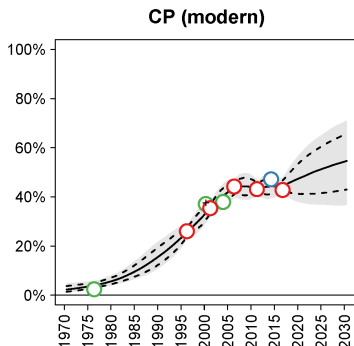
Introduction

- ▶ Motivating question:
How to estimate contraceptive prevalence using self-reported data collection, i.e. demographic health surveys (DHS)
- ▶ The current approach: Family Planning Estimation Model (FPEM, Cahill et al., 2018)

Ghana (Western Africa)



Nepal (Southern Asia)

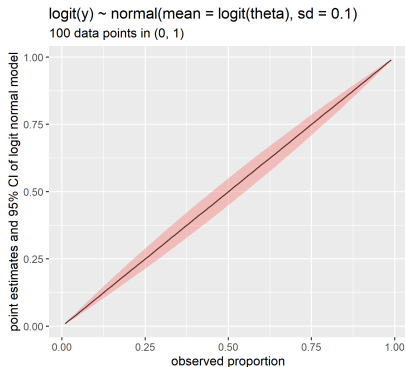


How is the data used in FPEM?

- ▶ FPEM: **data model** + process model
 - ▶ **data model**: describe how observed data relate to true prevalence

$$\text{logit}(y) \sim N(\text{logit}(\theta), \text{logit}.s^2 + \tau^2)$$

- ▶ Data model for modern contraceptive prevalence rate (mCPR)
 - ▶ y observed mCPR,
 - ▶ θ true mCPR,
 - ▶ $\text{logit}.s$ sampling error on logit-scale,
 - ▶ τ non-sampling error
- ▶ The visualization of data model assumption with $\text{logit}.s = 0, \tau = 0.1$

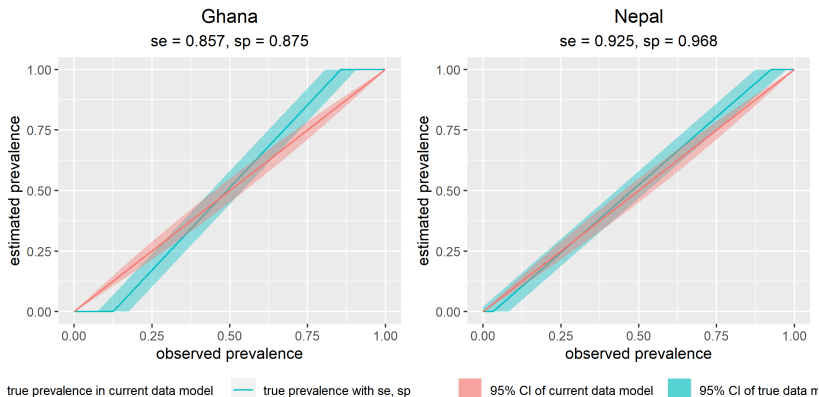


What if we observed some non-sampling error?

- ▶ Recap: no particular assumption of the non-sampling error
- ▶ BUT what if there is evidence of non-sampling error? e.g. observations of non-sampling error in the form of misclassification
- ▶ Two post-survey studies of DHS provides such evidence summarising into sensitivity se and specificity sp
 - ▶ 2014 Ghana DHS (Staveteig, 2017) (sample size = 48)
 $se = 0.857, sp = 0.875$
 - ▶ 2016 Nepal DHS (Staveteig et al., 2018) (sample size = 194)
 $se = 0.925, sp = 0.968$

The visualization of assumption based on misclassification

- ▶ The relationship between obs prev y and true prev θ
$$y = se \cdot \theta + (1 - sp)(1 - \theta)$$
- ▶ The true data model when subject to misclassification
$$y \sim N(se \cdot \theta + (1 - sp)(1 - \theta), s^2) T[0, 1]$$



How to do a better estimation

- ▶ Recap: current uncertainty assumption in FPEM is not OK with the potential existence of misclassification
- ▶ Question: **How to better estimate the uncertainty due to misclassification?**
- ▶ Motivation
 - ▶ Simple assumption of misclassification in the current assumption
 - ▶ Limited evidence of the misclassification from two post-survey studies

Our proposal: a new data model

- ▶ We propose a new data model to accomplish the mission
- ▶ The aims of the new model
 - ▶ $y = \theta$
 - ▶ If no misclassification: 95% CI determined by s
 - ▶ If misclassification: 95% CI determined by s, se^a, sp^a
- ▶ The form of the new model: a standardized likelihood function based on Normal-Laplace distribution (Reed, 2006)
 - ▶ $NL(\mu, \sigma, \alpha, \beta)$: the convolution of normal and Laplace distribution
 - ▶ The parameters of NL density is based on sampling error s , assumed misclassification se^a, sp^a
 - ▶ With a prior of $\theta \sim U(0, 1)$, the standardized likelihood is the posterior of θ given observed info

The visualization of the new data model

- ▶ We specify a parameterization routine based on the new model aims to pass the observed y , s and misclassification se^a , sp^a to the new data model
- ▶ Assume $s = 0.016$, $se^a = 0.9$, $sp^a = 1$
- ▶ Unchanged estimates, increased upper bound of 95% CI for estimates with increase of y
- ▶ plots to be added

Simulation of the new data model

- ▶ We fixed $\theta^{\text{true}} = 0.3, 0.5, 0.7$ and set various groups of true misclassification se, sp and assumed misclassification se^a, sp^a
- ▶ True data generation process with sample size n
true process formula tbd
- ▶ Compare posterior estimates from
 - ▶ true process model
 - ▶ Current assumption in FPEM (logit-normal)
 - ▶ The new proposed model (NL likelihood)
- ▶ The results shows that (1) the estimates is unchanged in the new NL model; (2) If misclassification exists: the new model did a better job than the old logit-normal model

New data model in FPEM

- ▶ New proposal vs the current data model in FPEM
- ▶ New model apply to DHS data points only as the evidence is from studies of DHS
- ▶ Allow user to define the misclassification $se^a = 0.9$, $sp^a = 1$
- ▶ Observe changes in the estimated trends and projection across countries
 - ▶ Red line/shades - new data model; Blue line/shades - old data model; Shapes: data points
- ▶ plots to be added

Conclusion

- ▶ We propose a new data model to account for increased asymmetric uncertainty associated with potential misclassification errors
- ▶ Coverage improvement in the simulation study when data are subject to misclassification
- ▶ Changes observed in FPEM implementation