#### Correcting for verbal autopsy misclassification bias in causespecific mortality estimates

Abhi Datta Department of Biostatistics Johns Hopkins University

#### Collaborators

JHU, **Biostatistics** 





Jacob Fiksel Brian Gilbert (former PhD students)



Sandipan Pramanik (Postdoctoral fellow)



Scott Zeger





Agbessi Amouzou

**Emily Wilson** 

Henry Kalter JHU, International Health



**Bob Black** 



Li Liu (JHU, PFRH)



Dianna Blau

(CHAMPS)



Ivalda Macicame (INS Mozambique)

- Funding (Bill and Melinda Gates Foundation):
  - Countrywide Mortality Surveillance for Action (COMSA) in Mozambique (PI: Amouzou)
  - CA-CODE: Child and Adolescent Causes of Death Estimation (PI: Liu)
  - Broadening the applicability of minimally-invasive-tissue-sampling (MITS)-based verbal autopsy (VA) calibration to improve global mortality estimates (PI: Datta)

### VA to COD to CSMF

COD = cause of death

CSMF = cause-specific mortality fractions

COMSA-Mozambique goal: Generate publicly available data on CSMFs at national and subnational levels using VA-records



### Minimally Invasive Tissue Sampling (MITS)

Data from CHAMPS project with both VA-COD and a minimally invasive tissue sampling (MITS)-COD

MITS-COD assignments been shown to be reasonably accurate when compared to the full diagnostic autopsies (Bassat et al. 2017)

CHAMPS data can be used to create a paired VA-MITS dataset to understand the accuracy of VA



MITS-VA Misclassification rates matrix

### Cause of death misclassification by VA



Misclassification rates of VA for neonates in COMSA-Mozambique

Single-cause MITS

Misclassification rates of VA for under-5 children in COMSA-Mozambique

This paired data reveals that VA misclassifies COD in a large % of deaths True for all age groups and choice of CCVA algorithm

## VA calibration

Law of total probability:  $p(a) = \sum p(a | y)p(y)$ 

y a = VA COD, y = MITS COD, M = [p(a=j|y=i)] is the misclassification rate matrix



# VA calibration

Law of total probability:  $p(a) = \sum p(a | y)p(y)$ 

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- Multinomial model for the unpaired COMSA VA data: estimates p(a)
- Multinomial model for the paired CHAMPS data: estimates the misclassification rates p(a|y)

# VA calibration

Law of total probability:  $p(a) = \sum p(a | y)p(y)$ 

• a = VA COD, y = MITS COD, M = [p(a=j|y=i)] is the misclassification rate matrix





- The true CSMF p(y) can be recovered consistently using the hierarchical model for the two data sources
- Hierarchical Bayesian modeling: Shrinkage prior to stabilize estimation of the misclassification rates

## **Ensemble VA calibration**

Many VA algorithms exist, each produces its own uncalibrated and calibrated CSMF

Ensemble calibration: Combines data from two (or more) VA algorithms into a single calibrated CSMF estimate.

Weights the algorithms in a datadriven way, favoring the more accurate ones.



## Multiple VA-COD

Many VA algorithms like InSilicoVA are probabilistic -- assigning scores to each cause being the underlying COD, i.e., compositional (or fractional) COD data

Create 10 copies of the VA record for each individual with causes of death assigned proportionately using the VA scores



All 10 copies will have the same MITS cause

Down-weight these copies by 1/10 in the calibration to keep the sample size same

# Multiple MITS-COD using generalized Bayes

- MITS diagnosis for many cases can also suggest multiple cause of death
  - Underlying and Immediate causes
- So both VA- and MITS-COD are compositional data
- Extend (weighted) multinomial likelihood to a Kullback-Liebler lossfunction (pseudolikelihood) for composition-on-composition regression
- Generalized Bayes: Using any loss function  $\ell(\theta | data)$  and a priori  $\pi(\theta)$
- ► Posterior:  $\Pi(\theta | data) \propto \exp(-\ell(\theta | data)) \Pi(\theta)$
- Unifies categorical and compositional COD data types

#### Overview of VA calibration pipeline for COMSA-Mozambique



### **COMSA-Mozambique neonatal results**

0

Calibrated

Uncalibrated



Model comparison of calibrated and uncalibrated CSMF using Widely Applicable Information Criterion WAIC

Uncalibrated

Calibrated

0

0

Calibrated

Uncalibrated

## Ongoing work

Transportability of MITS based VA-calibration for data from other countries

 Account for cross-site heterogeneity in misclassification rates

- Global application of VA-calibration for improved child mortality estimates (CA-CODE, Pi: Liu)
- VA-calibration using summary data
- Sub-group specific VA-calibration (downscaling)
- Use cause-hierarchy to calibrate at a finer cause-resolution

# Papers and software

- (Single-cause VA calibration) Datta, A., Fiksel, J., Amouzou, A., & Zeger, S. (2020). Regularized Bayesian transfer learning for population-level etiological distributions. Biostatistics, 1465–4644, 2020.
- (Multi-cause VA calibration) Fiksel, J., Datta, A., Amouzou, A., & Zeger, S. (2021). Generalized Bayes quantification learning under dataset shift. Journal of the American Statistical Association, 117(540), 2163-2181
- (Composition-on-composition regression) Fiksel, J., Zeger, S., & Datta, A. (2021). A transformation-free linear regression for compositional outcomes and predictors. Biometrics 78(3), 974–987, 2022.
- (Software) R-package calibratedVA for calibrating VA-based CSMF estimates using GBQL <u>https://github.com/jfiksel/CalibratedVA/</u>
- (Overview of VA-calibration) Fiksel J, et al. Correcting for verbal autopsy misclassification bias in causespecific mortality estimates American Journal of Tropical Medicine and Hygiene, 105(5S), 66-77, 2023.
- (Multi-cause VA-calibration for COMSA-Mozambique) Gilbert B et al. Multi-cause calibration of verbal autopsy-based cause specific mortality estimates of children and neonates in Mozambique American Journal of Tropical Medicine and Hygiene, 105(5S), 78-89, 2023.
- (COMSA-Mozambique main results) Macicame I et al. Countrywide Mortality Surveillance for Action in Mozambique: Results from a National Sample-Based Vital Statistics System for Mortality and Cause of Death American Journal of Tropical Medicine and Hygiene, 105(5S), 5-16, 2023.

Thank you



