Workshop & Algorithms

Python openVA Workshop Dar es Salaam, Tanzania August 8-10, 2023

Objectives

- Introduce Python openVA
- Train potential users
 - Inputs
 - Operation
 - Understanding outputs
- Obtain feedback that will help the openVA development team to improve Python openVA and related software

Thanks to our sponsors

- Vital Strategies
- CDC Foundation
- Bloomberg Philanthropies *Data for Health Initiative*
- Zambian Department of National Registration, Passport and Citizenship (DNRPC)

Organizers

- Vital Strategies Robert Mswia
- <u>openVA Team</u>
- Isaac Lyatuu
- Collins O'Chieng













Workshop overview

Day 1

- Morning: Introductions and verbal autopsy algorithms
- Afternoon: Python openVA Software

Day 2

- Morning: Worked examples using Python openVA, including hands-on work
- Afternoon 1: Interpreting openVA results
- Afternoon 2: Dashboard tools for verbal autopsy

Workshop overview

Day 3

- Morning: Supervised hands-on work using Python openVA on your own data
- Afternoon: Feedback and advice for future openVA development

Pacing & breaks

- First time we are doing this workshop
- Don't know how much time will be necessary for each section; therefore, no set times in schedule
- Breaks
 - Mid-morning coffee/tea: 10:30 10:50
 - Lunch: 1:00 2:00
 - Afternoon coffee/tea: 3:40 4:00

Introduction to verbal autopsy algorithms

- 1. Verbal autopsy algorithms
 - a. Physician-coded VA
 - b. Computer-coded VA
- 2. InterVA
- 3. InSilicoVA
- 4. Tariff

Verbal autopsy cause-coding algorithms

- Objectives of VA
 - Identify likely causes for individual deaths
 - Estimate cause-specific mortality fractions for population of deaths
- Components of VA
 - VA questionnaire
 - VA interview
 - VA data
 - Symptom-cause information (SCI)
 - Cause-classification logic
 - **Results: individual-level cause-specific metric (ICSM)** propensities, probabilities, tariff scores, etc.
 - Results: population-level cause-specific mortality fractions (CSMF)

Verbal autopsy cause classification

- This workshop focuses on automated cause-classification
- Cause classification **applies classification logic to VA data using SCI** to calculate a ICSM for each cause for each death and CSMFs
- Three elements required for classification
 - VA data
 - SCI
 - Classification logic
- All three elements affect the classification

Physician-coded VA - PCVA

- Physicians read VAs and identify likely causes
 - VA data
 - **SCI** is physicians' knowledge of relationships between VA indicators and causes
 - Logic is physicians' reasoning about VA data using their knowledge
- Can identify specific causes
- Time-consuming
- Potentially costly
- Usually slow
- Physicians often disagree \Rightarrow not replicable
- Not useful for large-scale VA

Computer-coded VA - CCVA

- Computer runs a statistical/computational algorithm
 - VA data
 - **SCI** is tabular data summarizing the relationship between VA indicators and causes
 - Logic is a formal statistical/mathematical/computational algorithm that combines VA data and SCI to produce ICSMs and CSMFs
- Less able to identify specific causes
- Very fast compared to physicians
- Effectively costless
- Highly replicable same or very nearly the same results every time algorithms are run
- Only realistic/feasible option for large-scale VA

- Interpret $VA \Rightarrow$ InterVA
- Developed by late Peter Byass and colleagues and refined over many years from roughly 2000 - 2020 (<u>DOI</u>)
- Supports standard WHO VAs: 2007, 2012, 2016
- Software available from Peter Byass' website: <u>http://www.byass.uk/interva/</u>
 - Windows installable and source code
 - No longer maintained
- Software available from openVA Team exactly replicated Byass' software
 - R package: <u>https://cran.r-project.org/web/packages/InterVA5/index.html</u> <u>continuously</u> <u>maintained</u>
 - Python openVA

- VA data
 - Requires WHO standard VA: 2007, 2012, 2016
- SCI
 - *Probbase*: conditional probabilities of observing a VA indicator given a specific cause
 - Elicited directly from physicians, **represents physicians' knowledge**
 - Quantifies relationship between each VA indicator and each cause in the WHO standard VA cause list
 - Does not quantify relationships between groups of causes and causes

- Logic
 - Inspired by Bayes' Rule but not mathematically probabilistic
 - **Only uses information on VA indicators that are present**; ignores information on VA indicators that are absent
 - For each death, produces individual-level propensity associated with each cause
 - Propensities with values less than 0.4 are ignored
 - Standard software reports top three propensities for each death as long when they exceed 0.4
 - **Produces undetermined causes:** if no propensity exceeds 0.4 for a given death, the cause is classified as *undetermined*
 - CSMFs calculated by summing up propensities for each cause across all deaths

- Considerations
 - Assumes VA indicators and causes related in a 1:1 fashion, ignores the effect of groups of VA indicators
 - Only utilizes information on VA indicators that were observed, ignores information associated with the fact that some VA indicators were not observed
 - Because of the 0.4 threshold for propensities, often identifies undetermined as the most likely 'cause' - typically about 20% of deaths
 - Algorithm is very simple \Rightarrow very fast to run
 - Raw VA data require checking and transforming into algorithm inputs this takes considerable time in all software
 - Part of openVA software

- InSilicoVA \Rightarrow *in-silicon* VA, like *in-vivo* and *in-vitro*, i.e. on a computer chip
- Created by <u>Zehang (Richard) Li</u>, <u>Tyler McCormick</u>, and <u>Sam Clark</u> (DOI) to improve on InterVA by:
 - Utilizing information on both present and absent VA indicators
 - Eliminating undetermined causes
 - Estimating uncertainty/confidence associated with both ICSMs and CSMFs
- Supports standard WHO VAs: 2012, 2016
- Software available from openVA Team
 - R package: <u>https://cran.r-project.org/web/packages/InSilicoVA/index.html</u> <u>continuously</u> <u>maintained</u>
 - Python openVA

- VA data
 - Requires WHO standard VA: 2012, 2016
- SCI
 - *Probbase*: conditional probabilities of observing a VA indicator given a specific cause
 - Elicited directly from physicians, **represents physicians' knowledge**
 - Quantifies relationship between each VA indicator and each cause in the WHO standard VA cause list
 - Does not quantify relationships between groups of causes and causes

- Logic
 - Mathematically valid probabilistic model
 - Estimates joint distribution of individual-level cause-specific probabilities and population-level cause-specific mortality fractions - this keeps individual and population levels consistent with each other
 - Uses full Bayes' Rule to calculate individual-level cause-specific probabilities
 - Utilizes information on both present and absent VA indicators
 - Produces distributions of cause-specific probabilities for each cause for each death
 - Produces distributions of cause-specific mortality fractions for a population of deaths
 - Distributions used to identify central probabilities and uncertainty/confidence
 - No undetermined causes
 - Easily identified causes have high probability and low uncertainty
 - Hard-to-identify causes have low probability and high uncertainty

- Considerations
 - Assumes VA indicators and causes related in a 1:1 fashion, ignores the effect of groups of VA indicators
 - Uses all information on both present and absent VA indicators
 - Does not produce undetermined causes
 - Produces information about uncertainty/confidence in ICSMs and CSMFs
 - Algorithm is very complex compared to InterVA \Rightarrow takes much longer to run
 - Raw VA data require checking and transforming into algorithm inputs **this takes considerable time in all software**
 - Part of openVA software

Tariff

- Created by Institute for Health Metrics and Evaluation (IHME)
- Uses only information from VA indicators that are present
- Supports Population Health Metrics Research Consortium (PHMRC) standard questionnaires: separate versions for *PHMRC long* and *PHMRC short*
- Software available from IHME: <u>SmartVA</u>

Tariff 2.0 implemented in SmartVA software

- VA data
 - Requires *PHRMC short* questionnaire
- SCI
 - SCI calculated as tariff values directly from '<u>PHMRC Gold Standard Dataset</u>' of reference deaths
 - Represents relationships between VA indicators and causes in PHRMC-short standard VA in about 10,000 reference deaths
 - Deaths from six facility locations in early 2000s *limited epidemiological context and* <u>not community deaths</u>
 - Does not quantify relationships between groups of causes and causes

Tariff 2.0 implemented in SmartVA software

• Logic

- Sums cause-specific tariff values for VA indicators that exist to produce tariff score for each cause; largest tariff score is most probably cause - <u>not mathematically probabilistic</u>
- **Only uses information on VA indicators that are present**; ignores information on VA indicators that are not present
- For each death, produces individual-level tariff score associated with each cause
- If no cause has large enough tariff score, cause is **undetermined**, similar to InterVA
- CSMFs calculated by summing individual tariff scores

Tariff 2.0 implemented in SmartVA software

- Considerations
 - Assumes VA indicators and causes related in a 1:1 fashion, ignores the effect of groups of VA indicators
 - Uses all information on both present and absent VA indicators
 - Produces undetermined causes
 - Algorithm is very simple \Rightarrow very fast to run
 - NOT part of openVA software

History of VA algorithms/software \Rightarrow Python openVA

- 1. WHO Standard verbal autopsy, InterVA, InSilicoVA, & openVA
- 2. PHMRC verbal autopsy and SmartVA

WHO standard VA, InterVA, InSilicoVA, & openVA

- WHO Verbal Autopsy Reference Group (VARG) responsible for determining and updated WHO VA standards
- First widely used standards is VA 2007, important contribution from INDEPTH standard VA
 - CCVA supported by InterVA3
- WHO 2012 standard
 - CCVA supported by InterVA4 and InSilicoVA
- WHO 2016 standard
 - Electronic data capture standard using Open Data Kit (**ODK**)
 - CCVA supported by InterVA5 and InSilicoVA
- WHO 2020
 - Standard instrument available in ODK *shorter and streamlined compared to 2016*
 - CCVA not yet supported; will be supported by InterVA5 and InSilicoVA within about one year

WHO standard VA, InterVA, InSilicoVA, & openVA

- Pre-openVA
 - Before InterVA5, InterVA was closed-source, proprietary software distributed free of charge as a Windows executable - *no one except Peter Byass really knew what it was doing*
 - Until recently SmartVA was also closed-source, proprietary software distributed as a Windows executable - *no one except IHME really knew what it was doing*
- openVA
 - Purpose: to make VA software <u>open source and free</u> so that <u>anyone can fully</u> <u>understand</u> what it is doing and use it
 - Reverse-engineered both InterVA and Tariff 1.0 and created open source R packages that implement the algorithms: <u>InterVA</u>, <u>Tariff 1.0</u>
 - Created and added InSilicoVA
 - Created R package openVA as a wrapper around the algorithms, including ability transform raw data and view/save results in various formats: <u>openVA</u>

openVA

• openVA Team

- Led by <u>Sam Clark</u>
- Software primarily <u>Richard Li</u> and <u>Jason Thomas</u>
- Many other members working on various aspects of VA
- openVA Pipeline
 - Uses Python, R, and SQL to automate retrieval of VA deaths from ODK, cause-classification, and pushing causes and raw VA data to DHIS-2
 - Command line and chron job
- All openVA software is open source
 - R CRAN, many packages
 - Github repositories at https://github.com/verbal-autopsy-software

Python openVA

- Context
 - Original openVA written using R and Java and distributed as R packages
 - All open source and free
 - R code utilizes myriad pre-existing packages
 - openVA R is a command line tool
 - Both openVA software and all packages and related software must be kept up to date OK for researchers but not production users
- Installing, maintaining, and running R-version of openVA is challenging/impossible for many users

Python openVA

- Python openVA
 - Aims to solve the useability issue facing many users of R openVA
 - Written exclusively in Python
 - Compiled into Windows and MacOS executables
 - Installs like a 'normal' application on both operating systems
 - \circ $\$ Has a windowed graphical user interface like a 'normal' application
 - User interacts with Python openVA by pointing and clicking **no command line**
 - Still free and open source source code (will be) available on Github repository after initial testing

PHMRC VA and SmartVA

• Tariff 1

- Uses PHMRC long questionnaire
- CCVA using old version of SmartVA
- Tariff 2
 - Uses PHMRC short questionnaire *much shorter*
 - CCVA supported by SmartVA Analyze software
 - Available as Windows and Linux executables
 - Also available on open-source Github repository
- openVA does not support PHMRC questionnaires (might in future)