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Correcting and estimating HIV mortality in Thailand based on 2005 verbal autopsy data focusing on demographic factors, 1996-2009

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Abstract

Background: It is known that death registry (DR) underestimates HIV deaths. The objectives of this study were to examine under-reporting/misclassification and to estimate HIV mortality in Thailand during 1996-2009 from a model based on 2005 verbal autopsy (VA) data.

Methods: Logistic regression was used to predict HIV deaths from the VA dataset with and without demographic covariates. This full model was then used to predict individual HIV deaths from the DR dataset of provinces in which VA was conducted. The proportions in the remaining provinces were predicted from spatial interpolation based on coefficients of the VA provinces.

Results: Area under Receiver Operating Characteristic curve of the full model was 0.969 compared to 0.879 of the simple cross-referencing model when demographic covariates were not included. DR-reported HIV deaths accounted for only one-third of all VA-estimated HIV deaths. The most misclassified HIV deaths were those registered as tuberculosis and mental and nervous system. Under-reporting was most common among females and people aged 20-39 years, and effect of province was highest in the upper north and upper south regions.

Conclusions: For approximately two-thirds of all HIV deaths estimated by the full model, the causes were reported under other categories, not HIV. Demographic variables are essential for accurately correcting causes of death from death registries.

Keywords: HIV death, Verbal autopsy, Death registry, Under-reporting, Misclassification, Estimation

Background

Inaccurate and unreliable attribution of causes of death is especially high where causes are ill-defined. Underreporting of HIV deaths is common in developing countries, and has been documented in Botswana [1], Brazil [2,3], South Africa [4,5], and Thailand [6-8], due to under-registered deaths and misclassification of cause of death. These problems severely limit the value of routine mortality data for public utility and affect resource allocations by policymakers.

Death registry (DR) data in Thailand provided by the Ministry of Interior through the Bureau of Policy and Strategy, Ministry of Public Health, are of poor quality, not only lacking completeness but also providing inaccurate cause of death [6,7,9-13]. Verbal autopsy (VA) surveys have been widely used in several countries including Uganda, China, Brazil, Tanzania, Bangladesh, South Africa, Zimbabwe, and Thailand to give more accurate information about causes of death [12,14,15]. The latest VA study in Thailand was carried out in 2005 in nine provinces by the SPICE (Setting Priorities using Information on Cost- Effectiveness analysis) project. The 2005 VA study [8,16-18] was used to estimate various causes of death including HIV. However, the simple cross-referencing method used in these studies ignored the effect of sex-age groups and locality of the deceased, which could give incorrect estimates due to confounding.



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We hypothesized that the utility of the 2005 VA data can be substantially improved if demographic variables were included to predict the cause of death. The aims of our study were 1) to examine under-reporting and misclassification of HIV deaths, based on modeling of the 2005 verbal autopsy data and 2) to estimate HIV deaths in all provinces of Thailand during 1996-2009.

Methods

Data sources and management

This study was confined to deaths of people aged 5 years and older, for which HIV death is common and often misclassified. DR data from 1996-2009 were obtained from the Bureau of Policy and Strategy database, Ministry of Public Health. The 2005 VA study was conducted by the SPICE project, and included a sample of 9,644 deaths (3,316 in-hospital and 6,328 outside-hospital) from 28 selected districts in nine provinces of four regions, of which 9,495 were deaths of persons aged 5 years and older.

Table 1 summarizes the cause groups based on VA counts. Accordingly, the chapter-block classifications of ICD-10 codes [19], consisting of blocks categorized mainly by human organs, were used to create 21 major cause groups for deaths at ages 5 years and older based on the

Table 1	Cause	aroups	based	on	VA	counts
	Cause	uloubs	Daseu		V A	counts

Cause groups	VA count	Percentages	
1: TB (A15-19)	195	2.1	
2: Septicemia (A40-41)	77	0.8	
3: HIV (B20-24)	512	5.4	
4: Other Infectious (A, B) ⁻	219	2.3	
5: Liver Cancer (C22)	500	5.3	
6: Lung Cancer ⁺ (C30-39)	320	3.4	
7: Other Digestive Cancer (C15-26 ⁻)	290	3.1	
8: Other Cancer (C ⁻ , D0-48)	697	7.3	
9: Endocrine (E)	647	6.8	
10: Mental, Nervous (F, G)	223	2.3	
11: Ischemic (I20-25)	617	6.5	
12: Stroke (I60-69)	1,076	11.3	
13: Other CVD (I ⁻)	540	5.7	
14: Respiratory (J)	801	8.4	
15: Digestive (K)	489	5.2	
16: Genitourinary (N)	412	4.3	
17: III-defined (R)	501	5.3	
18: Transport Accident (V)	536	5.6	
19: Other Injury (W, X0-59)	327	3.4	
20: Suicide (X60-84)	158	1.7	
21: All other	358	3.8	
Total	9,495	100.0	

⁺Respiratory/thoracic, ⁻exclude above.

distribution of VA-assessed deaths. For statistical accuracy, groups with small counts (mainly less than 200) were combined into larger groups using medical considerations (apart from septicemia, which received special attention due to over-reporting). The proportion of all deaths represented by these categories varied from 0.8% for septicemia to 11.3% for stroke and 5.4% for HIV deaths (ICD 10 code B20-24), as shown in Table 1.

Misclassification was not at random. The effects of sex, age, and spatial variables were used to correct misclassification, using logistic regression. For efficiency, the predictors were optimally grouped to obtain sufficient sample size for relatively homogeneous risk groups. Nine provinces were included in the VA study (Bangkok, Nakhon Nayok, Suphan Buri, Ubon Ratchathani, Loei, Phayao, Chiang Rai, Chumphon, and Songkhla). The effects of age for males and females were considered separately (see Results). Sex and age were grouped together into 14 levels (with seven levels of age in years: 5-19, 20-29, 30-39, 40-49, 50-59, 60-69, and 70+).

Similarly, misclassification of cause of death was considered differently for deaths in and outside hospitals. Reported causes of death and location were grouped into 18 levels, which resulted from the combination of two levels of location (in and outside hospital) and nine major causes of death (HIV, respiratory, septicemia, tuberculosis (TB), other infectious, mental and nervous system, digestive, ill-defined, and the remainder, which were aggregated into a single group).

Through logistic regression [20-22] we estimated the logit of the probability P that a person died from HIV as a linear function of the determinant factor. The simple logistic regression model with simple cross-referencing is formulated as

$$logit(P_i) = ln\left(\frac{P_i}{1 - P_i}\right) = \mu + \alpha_i, \tag{A}$$

where P_i is the probability of death due to HIV, μ is a constant, and α_i is the only parameter of DR causelocation *i*. The simple cross-referencing model (A) was compared with the full model (B), which includes an additive linear function of the determinant factors, which could be expressed as

$$logit(P_{ijk}) = ln\left(\frac{P_{ijk}}{1 - P_{ijk}}\right) = \mu + \alpha_i + \beta_j + \gamma_k, \qquad (B)$$

where P_{ijk} is the probability of death due to HIV and α_i , β_j and γ_k are individual parameters specifying DR cause-location group *i*, sex-age group *j* and province *k*, respectively.

We used "sum contrasts" developed by Tongkumchum and McNeil [23] and Kongchouy and Sampantarak [24] instead of conventional "treatment contrasts" where the first level is left out from the model to be the reference. This method allows us to compute the estimate and the 95% confidence interval of deaths for each of the covariate levels in the VA and the DR datasets.

To assess the accuracy of model prediction, the Receiver Operating Characteristic (ROC) curve from logistic regression was drawn based on a concept described by Chongsuvivatwong [22] and Fan et al. [25]. Area under the ROC curve (AUC) measures the performance of a model and represents model accuracy [26,27]. A cut-off point in the curve, where the predicted number of HIV deaths equals the observed value in the VA dataset (512 cases), was used to report sensitivity and specificity of the model. These were compared with results from the simple cross-referencing method.

Estimation of HIV mortality

For the nine study provinces, fitting the complete logistic regression model to the 2005 VA dataset resulted in nine province coefficients, 14 sex-age group coefficients, and 18 DR cause-location coefficients and the estimate of HIV deaths and 95% confidence intervals.

For the remaining 67 provinces, we used a simple and easily implemented spatial "triangulation method" [28,29] to interpolate province coefficients. This was preferred to the "kriging" method because it uses fewer points than kriging, and there were insufficient sample provinces (only nine) to provide the basis for kriging [30].

Triangles were drawn linking nine provinces in the 2005 VA study. The values of province coefficients in each triangle were assigned as an average of coefficients from nearby provinces in the model. For each triangle, values a, b and c were obtained by solving three equations using linear algebra based on latitude and longitude as follows.

$$a + long P_1 \times b + lat P_1 \times c = \beta_{P_1} \tag{C}$$

$$a + long P_2 \times b + lat P_2 \times c = \beta_{P2} \tag{D}$$

$$a + long P_3 \times b + lat P_3 \times c = \beta_{P3} \tag{E}$$

(Note: P = Province, β = coefficient)

The coefficient for any province *j* within a triangle could then be given by

$$\beta_{P_i} = a + long P_j \times b + lat P_j \times c \tag{F}$$

Coefficients for provinces outside triangles were obtained similarly by extrapolation from nearby provinces. Province coefficients for all provinces were thus obtained and the magnitude of HIV deaths estimated.

R program version 2.15.2 [31] was used for all statistical analysis and graphical displays.

Results

Ubon Ratchathani, Suphan Buri, and Chiang Rai had the largest numbers of total deaths (2373, 1600, and 1437 deaths, respectively), while Chumphon had the lowest (310 deaths). The VA-assessment gave 512 HIV deaths, whereas only 164 HIV deaths (32%) were correctly DR-reported.

From the likelihood ratio test in Table 2, the logistic regression model gives the deviance reduction between the full and null models as shown. All p-values are statistically significant.

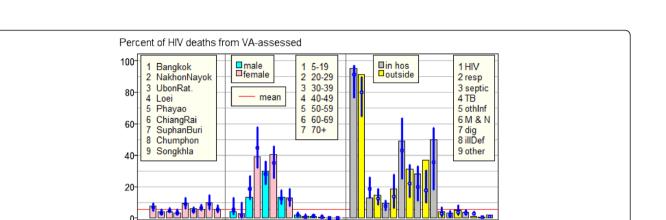
Figure 1 shows crude percentages of HIV deaths (among all deaths) by province, sex-age group, and DR causelocation group and the adjusted values with 95% confidence intervals. The values derived from the direct VA assessment and from the full model are similar, indicating variation among groups but with no substantial confounding. The plotted values above the average line reflect the groups that were more likely to die from HIV.

The 95% confidence interval for both Phayao and Chumphon is marginally higher than the mean, whereas for Loei it is marginally lower. Therefore, effect of province on misclassification of cause of death was marginal. The percentages of HIV deaths in age groups 20-49 are all substantially above the mean, with females higher than males when those aged 20-39 years were compared. Thus, age groups 20-49 were significantly more likely to have high levels of under-reporting. Finally, substantial numbers of HIV deaths were reported as TB, mental and nervous system, other infectious diseases, and respiratory for deaths in hospitals, whereas HIV deaths outside hospitals were reported as TB, other infectious diseases, and septicemia. These are the groups in which HIV deaths were often misclassified.

The full model was assessed using the ROC curve and compared with a simple cross-referencing model. Figure 2 shows the ROC curve of the simple crossreferencing model (model A) with only DR cause-location factor and the ROC curve of the full model (model B) with three factors of DR cause-location, sex-age groups, and province. The cut-off point marked by the star gives a total predicted number agreement of the number of VA-assessed HIV deaths in the model. The simple cross-referencing model represents an AUC of 0.879, 53.1% sensitivity, and 97.5% specificity, whereas the full model represents an AUC of 0.969, 69.3% sensitivity,

Table 2 Results from logistic regression, full model

Factors	Deviance reduction	df	p-value
DR cause-location	954.67	17	< 0.0000001
Sex-age group	783.58	13	< 0.0000001
Province	19.43	8	0.01271
Error	711.90	1154	



1 2 3

4 5 6

Figure 1 Estimated percentage of HIV-related deaths out of all deaths, based on the VA assessment (bars) and based on the full model (blue diamonds with error bars).

3 4 5 6

2

9

and 98.3% specificity. When we only compared an area above the diagonal line, the simple cross-referencing had AUC of 76% and the full model had AUC of 94%. In other words, the simple cross-referencing had an error of 24%, whereas the full model had an error of 6%. This means our model reduced the error by a factor of four. It is clear that the full model has the ability to predict the correct cause of HIV deaths better than the simple cross-referencing model. Just using the contingency table without statistical modeling, DR-reported cause had 32% sensitivity and 99.9% specificity.

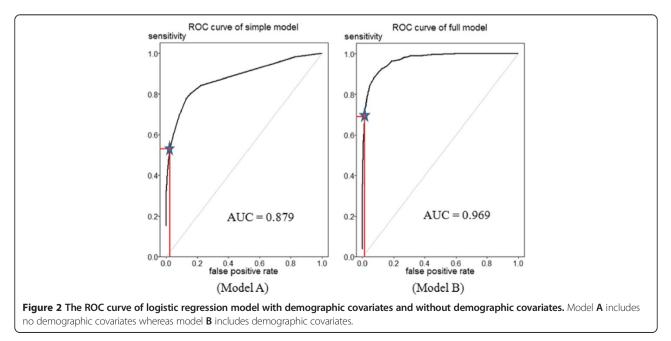
The full model (B) was then extended to all provinces. The left panel of Figure 3 shows the nine study province coefficients from the logistic regression model plotted in black. Values plotted in blue are averages of coefficients from nearby provinces in each triangle using the triangulation method. The right panel classifies province from the equations (*C*, *D*, *E*, *F* in methods section), according to three levels of coefficients. The highest were found in the upper north (Phayao and Phrae) and the upper south (Prachuap Khirikhan, Chumphon, Ranong, Surat Thani, and Phang Nga). This implies that HIV deaths were proportionally highest among all deaths in the upper north and the upper south.

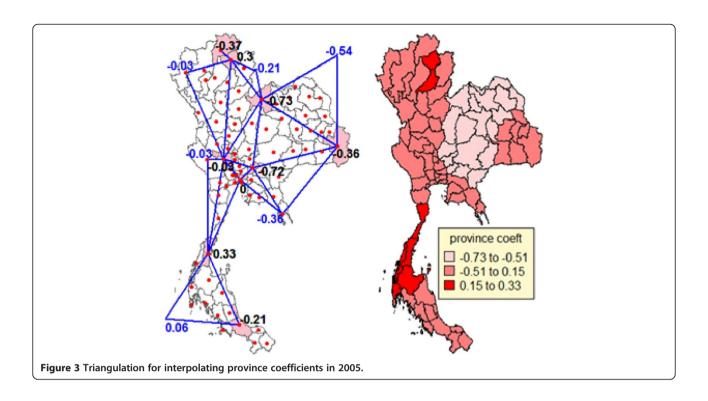
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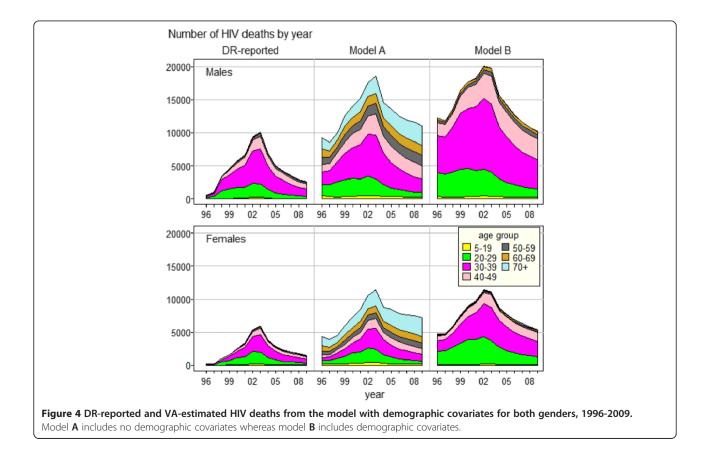
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Finally, the simple cross-referencing and the full model were then applied to the DR data for male and female deaths in 1996-2009 and plotted as area graphs in Figure 4. The area of each color strip denotes the number of HIV deaths in each age group. The total number of DR-reported HIV deaths were much lower than those estimated by cross-referencing (model A) and logistic regression (model B) by factors of 2.8 and 3.1, respectively.







While model A gave large proportions of HIV deaths at ages over 50 years (light blue, golden yellow, and grey), these were substantially reduced when the full logistic regression allowing for age/sex and province (model B) was used. On the other hand, for the young adult group, HIV cause of death was already substantially improved in accuracy by the simple logistic regression model.

Discussion and conclusions

Our logistic regression analysis showed that VA-assessed HIV deaths were more likely in female young adults compared to death registration, but many of those deaths were DR-registered as deaths from TB or from mental and nervous system disorders. A logistic regression-based method allowing for age/sex and geographical effects predicted HIV with higher sensitivity and specificity when compared with those HIV-estimated deaths derived from the cross-referencing from simple tabulation. Under-reporting was most common in the upper north and the upper south of the country. DR under-reported HIV deaths by a factor of three, whereas the simple cross-referencing method distorted the age distribution and could lead to a misunderstanding that HIV death was also common among the elderly.

HIV deaths were found to be relatively common among deaths in the age group 20-39 years, in agreement with other research [4,32]. AIDS is estimated to be the largest cause of death in Asian adults 15-44 years [33]. Before 1990, new HIV infections were highest among those injecting drugs and clients of sex workers. During 1995-2005, they were highest among the women with the category of housewife [34]. In other words, the most under-reporting of HIV deaths was found in females rather than males.

Most misclassifications of HIV deaths were classified as TB or mental and nervous system disorders. It is commonly known that TB and cryptococcal meningitis are the leading causes of opportunistic infections among HIV patients [35-38]. These infections were possibly recorded as the primary cause of deaths in death certificates either to avoid stigma to the family of the deceased, because the symptoms of TB and HIV are very similar, or because the people reporting the death might not have access to the results of a HIV test for the deceased. Another general condition often recorded was "immunodeficiency (D849: immunodeficiency, unspecified)." This might in fact be the more specific "HIV/AIDS (B20-B24: Human immunodeficiency virus disease)" in ICD10 coding [7].

Misclassification was associated with region (province). This could be due to difference in the levels of intensity of the HIV epidemic, stigmatization, and availability of qualified personnel for DR recording and their attitude toward HIV-related death across the regions. HIV mortality peaked in the upper north, especially in Phayao, because in the past two decades the HIV epidemic has been most severe in the upper north [32,34,39-42]. Onethird of HIV deaths were predicted in the northern region since 1987-2014 [39]. Those HIV deaths were higher in the upper south than in the central region in spite of the less severe HIV epidemic [40]. HIV deaths in the south were more likely to be misclassified to other causes, as the area was perceived to have low levels of HIV [43]. In addition, mortality varies by geographic location, and the south has the lowest overall mortality [43,44].

Our full logistic regression model based on the 2005 VA data was shown to predict and estimate HIV deaths with high sensitivity, specificity, and AUC, better than the simple cross-referencing model. The specificity level from our model was higher than a verbal autopsy tool from Uganda, where sensitivity was not reported [45]. The cross-referencing method has been used in many previous studies [1,15-18,45]. Inadequate models can give misleading or incorrect inferences [26]. Our study showed that the use of this simple method should be discouraged because it distorts the HIV death estimate in various demographic groups. This distortion can mislead priority setting and resource allocation.

There were limitations in our analysis. First, the sample survey design did not stratify by strong predictors of the outcome such as reported cause and location of report. The study sample thus did not adequately cover the population at risk for HIV and the sample size did not allow precise estimation among certain minority groups, such as the Muslim group in the far south. Second, only nine of Thailand's 76 provinces were included in the VA study.

Third, we have assumed that the 2005 VA data can inform corrections in all years between 1996 and 2009, while it is clear that the coverage of antiretroviral treatment was near zero in 1996, 12% in 2003, 41% in 2005, and 76% in 2009 [46]. There would therefore be differences in misclassification of HIV-related deaths across the years, which are not captured by our methods. Finally, VA itself has limitations, in terms of inaccuracy of informants and recall bias. The results must therefore be carefully interpreted.

Competing interests

The authors declare that they have no competing interests.

Authors' contributions

All authors participated in the design of the study and the interpretation of the results. AC and PT performed the study analysis and drafted the manuscript and prepared all tables and graphs. VC critically reviewed the manuscript. All authors read and contributed to the final manuscript.

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