## Original Article

# Validation of HIV risk screening tool to identify infected adults and adolescents greater than 14 years at community level in Tanzania and Zambia 

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#### Abstract

Introduction: There are several risk factors being used to identify undiagnosed HIV-infected adults. As the number of undiagnosed people gets less and less, it is important to know if existing risk factors and risk assessment tools are valid for use. Methods: Data from the Tanzania and Zambia Population-Based HIV Impact Assessment (PHIA) household surveys which were conducted during 2016 was used. We first included 12 risk factors (being divorced, separated or widowed; having an HIV+ spouse; having one of the following within 12-months of the survey: paid work, slept away from home for $\geq 1$-month, having multiple sexual partners, clients of sex workers, sexually transmitted infection, being tuberculosis suspect, being very sick for $\geq 3$-months; ever sold sex; diagnosed with cervical cancer; and had TB disease into a risk assessment tool and assessed its validity by comparing it against HIV test result. Sensitivity, specificity and predictive value of the tool were assessed. Receiver Operating Characteristic (ROC) curve comparison statistics was also used to determine which risk assessment tool was better. Results: HIV prevalence was $2.3 \%(2.0 \%-2.6 \%)$ ( $n=14,820$ ). For the tool containing all risk factors, HIV prevalence was $1.0 \%$ when none of the risk factors were present (Score 0) compared to $3.2 \%$ when at least one factor (Score $\geq 1$ ) was present and $8.0 \%$ when $\geq 4$ risk factors were present. Sensitivity, specificity, PPV, and NPV were $82.3 \%(78.6 \%-85.9 \%), 41.9 \%(41.1 \%-42.7 \%)$, $3.2 \%(2.8 \%-3.6 \%)$, and $99.0 \%(98.8 \%-99.3 \%)$, respectively. The use of a tool containing conventional risk factors (all except those related with working and sleeping away) was found to have higher AUC ( 0.65 vs 0.61 ) compared to the use of all risk factors ( $p$ value $<0.001$ ). Conclusions: The use of a screening tool containing conventional risk factors improved HIV testing yield compared to doing universal testing. Prioritizing people who fulfill multiple risk factors should be explored further to improve HIV testing yield.


Keywords: Adult HIV risk assessment tool; Undiagnosed HIV; Never tested for HIV; HIV testing yield, Tanzania, Zambia
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## Introduction

HIV testing is the gateway for case finding, care, and treatment, as well as prevention services for high-risk individuals (1-3). Over the years, remarkable progress has been made in diagnosing infected people and putting them on treatment. To date, Eastern and Southern African countries have coverage of $87 \%$ ( $77 \%-95 \%$ ) compared to the $90 \%$ UNAID's target of diagnosing $90 \%$ of all estimated HIV infected people by 2020 , while the coverage is $68 \%(54 \%-87 \%)$ for Central and Western African countries (4). This correlates with the high uptake of HIV testing across these countries. Prior HIV testing among surveyed men and women $15-49$ years was $62 \%$ and $74 \%$, respectively for Eastern, Southern, and Central African
countries, respectively, from 2015-2018. It was much lower for Western African countries at $31 \%$ for women and $16 \%$ for men $(5,6)$.
Maintaining such high testing coverage or conducting door-to-door testing in high-risk communities is not feasible because of the limited funding available for HIV programs considering the flattening of global support for HIV programs, especially that of PEPFAR over the past 10 years $(7,8)$. As a result of that, a strategic shift has been made to implement targeted HIV testing with the aim of getting high testing yield per dollar spent on HIV test kit in many country HIV programs, including highburden countries to put as many infected people on treatment and reduce new infection and mortality
in the process (9).
A number of HIV risk factors have been identified and in use to effect targeted HIV testing of at-risk people. The World Health Organization (WHO) recommends HIV testing for clients having sexually transmitted infection (STI), viral hepatitis, tuberculosis (TB); key populations including commercial sex workers, men having sex with men, and IV drug users; clients with symptoms or medical conditions that could indicate HIV infection, including presumed and confirmed TB cases (1, 2). Other factors known to increase the risk of HIV infection include having multiple sexual partners (10, 11), being divorced, separated or widowed (DSW) (12), history of being a client of a sex worker (13), having cervical Cancer (14), being partners with known infected person (15). A number of HIV risk assessment tools were validated in different settings in an effort to determine best options to identify HIV infected adults (10, 16,17). These tools often don't include risk factors recommended by the WHO and in use in high prevalence countries. Knowing the performance and limitation of a risk screening tool containing all common HIV risk factors is crucial to determine case finding strategies that better fit routine implementation setting and assess quality of testing services both in clinical and community settings. This study aims to:

1. determine the performance of a hypothetical HIV risk assessment tool that contains conventional HIV risk factors to identify undiagnosed HIV + adults and adolescents $>14$ years,
2. determine the performance of a hypothetical HIV risk assessment tool that contains all potential HIV risk factors to identify undiagnosed HIV+ adults and adolescents >14 years,
3. determine the performance of a hypothetical HIV risk assessment tool that contains only statistically significant HIV risk factors to identify undiagnosed HIV positive adults and adolescents $>14$ years
4. determine which of the above three tools is better in terms of overall performance to identify undiagnosed HIV positive adults and adolescents $>14$ years
5. determine if the presence of multiple HIV risk factors in one person improves performance of risk assessment tool to identify undiagnosed HIV positive adults and adolescents $>14$ years.

## Materials and Methods

Study setting and design: This is a cross-sectional study based on secondary analysis of data from two community-based household surveys that were conducted in Zambia (2016) and Tanzania (2016-2017). These surveys were Population-Based HIV Impact

Assessment (PHIA) studies conducted with PEPFAR support. These study countries are high burden countries providing donor driven (PEPFAR and Global Fund) HIV program. HIV response is mostly facility based where diagnostic and treatment services are provided with community component mainly focusing on prevention and care aspect. PHIA surveys are cross-sectional nationally representative, householdbased surveys. The main objectives of the surveys are to measure national HIV prevalence and viral suppression, nationally and sub-nationally, to assess the impact of HIV treatment and prevention programs in each country. Participants were selected using twostage, stratified cluster sampling. Participants who provided written consent undergo interview based on a structured questionnaire and a biomarker for HIV testing $(18,19)$.
Inclusion criteria: adolescents and adults $>14$ years who had matching interview and biomarker datasets (HIV testing result) and who had never tested for HIV prior the survey was included.
Sample size: was calculated to allow comparison between areas under receiver operating characteristic (ROC) curves between two different risk assessment tools. Sample size was calculated to be 1,363 assuming $\mathrm{AUC} 1=0.65$, $\mathrm{AUC} 2=0.6$, alpha $=0.05$, power $=$ $80 \%$, correlation in positive group $=0.4$, and correlation in negative group $=0.4$ (21).
HIV risk factors: the following variables were considered in different HIV risk assessment tools to generate a tool with better sensitivity, specificity, and positive predictive value ( $\mathrm{PPV}+$ ): being divorced, separated or widowed (DSW) (12), having an HIV + spouse (15), having paid work within 12 months of the survey, slept away from home for at least a month within 12 months of the survey, having multiple sexual partners $(10,11)$ within 12 months of the survey, had ever sold sex (13),clients of sex workers within 12 months of the survey, had sexually transmitted infection (STI) within 12 months of the survey $(1,2)$, diagnosed with cervical cancer (14), being a tuberculosis (TB) suspect within 12 months of the survey which meant having any of the following symptoms: cough, fever, night sweats or weight loss, had TB disease, past or present, and being very sick for at least 3 months within 12 months of the survey, that is being too sick to work or do normal activities (1,2).

## HIV risk assessment tools examined:

Four different hypothetical tools were considered in the validation:

Tool 1: A tool that contained all conventional and any newly identified statistically significant risk factors that predicted HIV infection status in individuals never tested for HIV,
Tool 2: A tool that contained only statistically significant risk factors associated HIV infection,
Tool 3: A tool that contained conventional
risk factors only, and
Tool 4: A tool that contained conventional risk factors and a combination of newly identified risk factors.
HIV testing: was offered for everyone in the survey and performed for all consenting adults and adolescents $>14$ years during the survey. Known HIV+ status was further confirmed through the use of antiretroviral markers within the blood. Those with antiretroviral markers were excluded from the study.

## Data analysis

Data were obtained from the public domain of PHIA website (22) and analyzed using Stata 14.0 statistical software. First, risk factors that had association with HIV infection among those who never tested for HIV were identified using Chi Square test. To develop scores for a risk assessment tool, appropriate screening items were selected and coded one when the risk factor was present and zero when it was not and the total score calculated for each individual as the sum of the numerical values of the screening items included within a tool. For instance, for the first screening tool where all risk factors were included, the minimum score was 0 while the potential maximum was 12.Chi Square test was also done to examine if having risk screening score of $\geq 1, \geq 2, \geq 3$, or $\geq 4$ was associated with HIV infection. Sampling weights were used to adjust statistical values taking into account complex sampling design used in PHIA surveys (20). To determine the optimal cut-off for the screening tool that will enable identification of people at risk of HIV infection, a receiver operating characteristic (ROC) curve was plotted. The area under the ROC (Receiver Operating Characteristic) curve, Area Under the Curve (AUC) and corresponding sensitivity, specificity, positive predictive (PPV) and negative predictive values (NPV) using the screening tool at different levels of scores were determined. ROC comparison statistics was used to statistically test equality between AUC of the different scores. For the score selected to be having the best combination of sensitivity, specificity, PPV, NPV and AUC, similar analysis was conducted to see if age, gender, and residence affected AUC. This was done by doing stratified analysis of AUC using the stated variables.
Finally, to compare and select between the different risk assessment tools, test of quality of AUC was done. Number needed to test to identify one HIV infected person (NNT + ) was also calculated to see if risk assessment tools reduced this number compared to universal testing. To select appropriate variables for the second tool, logistic regression was used. Variables with p- value $<0.25$ during bi-variable analysis were included in the final model. Variables with pvalue $<0.05$ in the final model were included in the making of Tool 2.
Ethical considerations: Both PHIA surveys had written informed consent, both for interview and blood collection for all participating adults. Parents
consented for adolescents. All datasets don't have individual identifiers like names or addresses that can be used to identify people. In addition, the study got a non-human subject determination from the Office of International Research Ethics of Family Health International (FHI360).

## Results

## Selection of study participants

Of 68,564 adults and adolescents $>14$ years included in the Tanzania and Zambia PHIA studies, 55,340 had a matching interview and biomarker (including HIV testing) datasets. Of these, 39,103 (70.7\%), had previously been tested for HIV and 181 ( $0.3 \%$ ) had missing previous HIV testing data, and hence excluded from the analysis. Of the remaining $16,056,15160$ ( $94.4 \%$ ) were tested for HIV. Excluding those who didn't have sampling weights, the final study sample was 14,820 . (Figure 1)

Compared to individuals who were eligible for inclusion in this study but had missing HIV testing result, those who were tested for HIV during the survey were likely to be older, male, from urban area. They were also less likely to have multiple sexual partners or sexually transmitted infection in the past 12 months. (Table 1)


Figure 1.Selection of Study Sample

Table 1. Comparison between Study Participants and non-Participants based on Socio-demographic Characteristics and HIV Risk Factors

| Variable | Response | Total, $\mathbf{n}$ | Study Participants (Tested for HIV during survey), n (\%) | Non-participants (Not tested for HIV during Survey), n (\%) | $P$ value |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Age | 15-24 | 8,297 | 7868 (53.1\%) | 429 (47.9\%) | $<0.001$ |
|  | 25-49 | 3,789 | 3507 (23.7\%) | 282 (31.5\%) |  |
|  | 50+ | 3,630 | 3445 (23.2\%) | 185 (20.6\%) |  |
| Gender | Male | 8,488 | 7945 (53.6\%) | 543 (60.6\%) | $<0.001$ |
|  | Female | 7,228 | 6875 (46.4\%) | 353 (39.4\%) |  |
| Residence | Urban | 4,817 | 4461 (30.1\%) | 356 (39.7\%) | $<0.001$ |
|  | Rural | 10,899 | 10359 (69.9\%) | 540 (60.3\%) |  |
| Education | No education | 2,887 | 2753 (18.6\%) | 134 (15.0\%) | $<0.001$ |
|  | Primary | 8,243 | 7792 (52.6\%) | 451 (50.3\%) |  |
|  | Secondary | 4,372 | 4096 (27.6\%) | 276 (30.8\%) |  |
|  | Tertiary | 214 | 179 (1.2\%) | 35 (3.9\%) |  |
| Wealth Quintile | Lowest | 3,785 | 3587 (24.2\%) | 198 (22.1\%) | $<0.001$ |
|  | Secondary | 3,613 | 3442 (23.2\%) | 171 (19.1\%) |  |
|  | Middle | 3,336 | 3168 (21.4\%) | 168 (18.8\%) |  |
|  | Fourth | 2,460 | 2291 (15.5\%) | 169 (18.9\%) |  |
|  | Highest | 2,522 | 2332 (15.7\%) | 190 (21.2\%) |  |
| Marital status | Single, Married | 13,772 | 12981 (87.6\%) | 791 (88.3\%) | 0.542 |
|  | DSW ${ }^{\text {s }}$ | 1,944 | 1839 (12.4\%) | 105 (11.7\%) |  |
| Spouse is Known to have HIV | No | 15,673 | 14777 (99.7\%) | 896 (100.0\%) | 0.106 |
|  | Yes | 43 | 43 (0.3\%) | 0 (0.0\%) |  |
| Having Paid Work* | No | 10,842 | 10212 (68.9\%) | 630 (70.3\%) | 0.377 |
|  | Yes | 4,874 | 4608 (31.1\%) | 266 (29.7\%) |  |
| Slept Away from Home for $>1$ month* | No | 13,708 | 12916 (87.2\%) | 792 (88.4\%) | 0.280 |
|  | Yes | 2,008 | 1904 (12.8\%) | 104 (11.6\%) |  |
| Multiple Sexual Partners* | No | 14,125 | 13292 (89.7\%) | 833 (93.0\%) | 0.002 |
|  | Yes | 1,591 | 1528 (10.3\%) | 63 (7.0\%) |  |
| Ever Sold Sex | No | 14,125 | 13292 (89.7\%) | 833 (93\%) | 0.428 |
|  | Yes | 1,591 | 1528 (10.3\%) | 63 (7.0\%) |  |
| Clients of Sex Workers* | No | 14,962 | 14107 (95.2\%) | 855 (95.4\%) | 0.749 |
|  | Yes | 754 | 713 (4.8\%) | 41 (4.6\%) |  |


| Variable | Response | $\begin{aligned} & \text { Total, } \\ & \text { n } \end{aligned}$ | Study Participants (Tested for HIV during survey), $n$ (\%) | Non-participants (Not tested for HIV during Survey), n (\%) | $\begin{aligned} & \hline \mathbf{P} \\ & \text { value } \end{aligned}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Sexually transmitted infection* | No | 14,382 | 13522 (91.2\%) | 860 (96.0\%) | $\begin{array}{r} <0.00 \\ 1 \end{array}$ |
|  | Yes | 1,334 | 1298 (8.8\%) | 36 (4.0\%) |  |
| Has Cervical Cancer | No | 15,632 | 14741 (99.5\%) | 891 (99.4\%) | 0.921 |
|  | Yes | 84 | 79 (0.5\%) | 5 (0.6\%) |  |
| Presumptive TB ${ }^{\# *}$ | No | 15,356 | 14479 (97.7\%) | 877 (97.9\%) | 0.726 |
|  | Yes | 360 | 341 (2.3\%) | 19 (2.1\%) |  |
| TB disease, current or past | No | 15,547 | 14657 (98.9\%) | 890 (99.3\%) | 0.225 |
|  | Yes | 169 | 163 (1.1\%) | 6 (0.7\%) |  |
| Sick for the past 3 months* | No | 15,073 | 14214 (95.9\%) | 859 (95.9\%) | 0.953 |
|  | Yes | 643 | 606 (4.1\%) | 37 (4.1\%) |  |
| Total |  | 15,716 | 896 (100\%) | 14,820 (100\%) |  |

${ }^{\$}$ Divorced, Separated, or Widowed, *within the last 12 months of the survey,
\#Cough, fever, night sweats, or weight loss

## Characteristics of study participants

Of 14,820 study participants, $57.8 \%$ were men, and had a median age of 30 (IQR: 21-24). HIV prevalence was $2.3 \%$ ( $95 \%$ (CI): 2.0-2.6). HIV prevalence was
higher for the age category $25-49$, among women, and in urban settings, while it was lowest for those with secondary education and those in the highest wealth quintile. (Table 2)

Table 2 Socio-demographic Factors Associated with HIV Infection among adults and adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia (2016) and Tanzania (2016-2017)

| Variable | Response | Total, $\mathbf{n}$ | HIV+, n (\%) | P value |
| :--- | :--- | ---: | ---: | ---: |
| Age | $15-24$ | 8,114 | $49(0.6 \%)$ | $<\mathbf{0 . 0 0 1}$ |
|  | $25-49$ | 3,443 | $196(5.7 \%)$ |  |
| Gender | $50+$ | 3,263 | $95(2.9 \%)$ |  |
|  | Male | 8,567 | $162(1.9 \%)$ | $\mathbf{0 . 0 0 2}$ |
| Residence | Female | 6,253 | $176(2.8 \%)$ |  |
| Education | Urban | 4,707 | $130(2.8 \%)$ | $\mathbf{0 . 0 2 6}$ |
|  | Rural | 10,113 | $207(2.1 \%)$ |  |
|  | No education | 2,647 | $82(3.1 \%)$ | $\mathbf{0 . 0 0 1}$ |
|  | Primary | 8,182 | $197(2.4 \%)$ |  |
| Wealth Quintile | Secondary | 3,859 | $55(1.4 \%)$ |  |
|  | Tertiary | 132 | $4(3.1 \%)$ |  |
|  | Lowest | 3,398 | $73(2.1 \%)$ | $\mathbf{0 . 0 3 8}$ |
|  | Secondary | 3,412 | $70(2.0 \%)$ |  |
|  | Middle | 3,081 | $73(2.4 \%)$ |  |
|  | Fourth | 2,432 | $79(3.2 \%)$ |  |
| Total | 2,496 | $43(1.7 \%)$ |  |  |

[^0] weight loss

## Statistically significant predictors of HIV infection

Table 3 summarizes predictors of HIV infection by including only factors being examined in the development of factors included in the development of risk assessment tool. Accordingly, all factors were predic-
tors of infection except for the following variables: Presumptive TB, TB disease, and being sick in the past 3 months.

Table 3. Predictors of HIV Infection among adults and adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia (2016) and Tanzania (2016-2017)

| Variable | Response | Total, n | $\begin{gathered} \text { HIV+, n } \\ (\%) \end{gathered}$ | Crude Odds Ratio | $\mathbf{P}$ value | Adjusted Odds Ratio | $P$ value |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Marital status | Single, <br> Married | 13,045 | 230 (1.8\%) | 1 |  | 1 |  |
|  | DSW ${ }^{\text {s }}$ | 1,775 | 108 (6.1\%) | 3.6 (2.7-4.8) | $<0.001$ | 3.7 (2.7-5) | $<0.001$ |
| Spouse is Known to have HIV | No | 14,777 | 333 (2.3\%) | 1 |  | 1 |  |
|  | Yes | 43 | 5 (11.0\%) | 5.3 (1.8-15.6) | 0.004 | 5.7 (1.8-18.2) | 0.005 |
| Having Paid Work* | No | 9,692 | 172 (1.8\%) | 1 |  | 1 |  |
|  | Yes | 5,128 | 166 (3.2\%) | 1.8 (1.4-2.4) | $<0.001$ | 1.7 (1.3-2.2) | 0.001 |
| Slept Away from Home for >1 month* | No | 12,939 | 273 (2.1\%) | 1 |  | 1 |  |
|  | Yes | 1,881 | 65 (3.5\%) | 1.7 (1.1-2.5) | 0.016 | 1.4 (0.9-2.2) | 0.089 |
| Multiple Sexual Partners* | No | 12,970 | 282 (2.2\%) | 1 |  | 1 |  |
|  | Yes | 1,850 | 58 (3.1\%) | 1.4 (1.0-2.0) | 0.042 | 1 (0.6-1.5) | 0.889 |
| Ever Sold Sex | No | 14,786 | 333 (2.3\%) | 1 |  | 1 |  |
|  | Yes | 34 | 5 (13.5\%) | 6.8 (3.2-14.5) | $<0.001$ | 6.8 (3.1-14.9) | $<0.001$ |
| Clients of Sex Workers* | No | 14,100 | 304 (2.2\%) | 1 |  | 1 |  |
|  | Yes | 720 | 36 (4.9\%) | 2.3 (1.5-3.5) | $<0.001$ | 1.5 (0.9-2.5) | 0.127 |
| Sexually transmitted infection* | No | 13,385 | 274 (2.0\%) | 1 |  | 1 |  |
|  | Yes | 1,435 | 65 (4.5\%) | 2.3 (1.6-3.2) | $<0.001$ | 1.9 (1.3-2.9) | 0.004 |
| Has Cervical Cancer | No | 14,755 | 335 (2.3\%) | 1 |  | 1 |  |
|  | Yes | 65 | 4 (5.5\%) | 2.6 (1.1-5.9) | 0.029 | 2 (0.8-5.1) | 0.152 |
| Presumptive TB ${ }^{\# *}$ | No | 14,460 | 326 (2.3\%) | 1 |  | 1 |  |
|  | Yes | 360 | 12 (3.3\%) | 1.5 (0.8-2.8) | 0.231 | 0.7 (0.2-2.6) | 0.616 |
| TB disease, current or past | No | 14,639 | 330 (2.3\%) | 1 |  | 1 |  |
|  | Yes | 181 | 8 (4.4\%) | 2.0 (0.9-4.2) | 0.07 | 2.2 (0.5-9.1) | 0.264 |


| Variable | Response | Total, $\mathbf{n}$ | HIV,$+ \mathbf{n}$ <br> $\mathbf{( \% )}$ | Crude <br> Odds Ratio | P value | Adjusted <br> Odds Ratio | P value |
| :--- | :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| Sick for the past 3 <br> months* | No | 14,242 | $317(2.2 \%)$ | 1 |  | 1 |  |
|  |  | 578 | $22(3.8 \%)$ | $1.7(0.9-3.1)$ | 0.086 | $1.4(0.7-2.6)$ | 0.347 |
| Total | Yes | 14,820 | $338(2.3 \%)$ |  |  |  |  |

${ }^{\$}$ Divorced, Separated, or Widowed; *within the last 12 months of the survey, ${ }^{\#}$ Cough, fever, night sweats, or weight loss

## HIV Prevalence by Risk Factors

Figure 2 summarizes HIV prevalence by risk factor. The highest was recorded for people who sold sex (13.5\%), followed by spouses of HIV-infected adults (11\%) and those who were divorced, separated, or
widowed (6.1\%). The presence of other risk factors had HIV testing yield ranging from 3.1\%-5.5\%. TB in the past 10 years had a testing yield of $33.3 \%$ for Zambia compared to $3.7 \%$ for those without TB in the past 10 years, p value $<0.001$ (data not shown).


Figure 2. HIV Testing Yield by Risk Factors among Adults and Adolescents $>14$ years who were never tested

## Determining cut off (risk score) for risk assessment tools

Looking at the performance of Tool 1 at different risk score levels, those individuals having one or more risk factors were found to have an HIV prevalence of $3.2 \%$ which increased with increasing cut-off: $4.4 \%$, $5.6 \%, 7.9 \%$ HIV prevalence for two, three, and four cut-off scores respectively (Table 4). Area under the curve (AUC) can be seen to reduce as the risk assess-
ment cut-off increases. A score of $\geq 1$ was found to have the highest sensitivity at $82.3 \%$ ( $95 \% \mathrm{CI}$ : 78.6\% $-85.9 \%$ ) with the next score of $\geq 2$ having nearly half the sensitivity at $46.8 \%(42.0 \%-51.6 \%)$. The specificity was higher for a higher cut-off. Positive predictive value was higher for a higher cut-off point while negative predictive value was comparable between all cut-off scores (Figure 3).

Table 4. Association of HIV Risk Scores with HIV Infection using a tool that contains all HIV Risk Factors for Adults and Adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia (2016) and Tanzania (2016-2017)

Figure 3. Receiver Operating Characteristics Curve by HIV Risk Scores for Adults and Adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia (2016) and Tanzania (2016-2017).

The sensitivity, specificity, and area under the curve are indicated for each cut-off. Compared to a cut-off score of $\geq 1$, AUC was comparable with a cut-off score of $\geq 2$, while it was lower for those with higher cut-off scores ( p value $<0.001$ ). (Table 5) The AUC was comparable by age, gender, and residence. (Table 6)

Table 5. Sensitivity, Specificity, Positive Predictive Value (PPV+), and Negative Predictive Value (NPV) for HIV Risk Screening Tool Containing Various Combinations of All Risk Factors

* Score $\geq 1$ means an individual who has one or more risk factors for HIV; ** AUC= Area under the Curve of a Receiver Operating Curve;

| Screening <br> Tool Score* | Sensitivity | Specificity | PPV | NPV | $\underset{* *}{\text { AUC }}$ | $\mathbf{P}$ <br> Value *** |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Score $\geq 1$ |  |  |  |  |  |  |
|  | 82.3\% | 41.9\% | 3.2\% | 99.0\% |  |  |
|  | (78.6\%- | (41.1\%- | (2.8\%- | (98.8\%- |  |  |
|  | 85.9\%) | 42.7\%) | $3.6 \%)$ | 99.3\%) | 0.612 |  |
| Score $\geq 2$ |  |  |  |  |  |  |
|  | 46.8\% | 76.4\% | 4.4\% | 98.4\% |  |  |
|  | (42.0\%- | (75.7\%- | (3.7\%- | (98.2\%- |  |  |
|  | 51.6\%) | 77.1\%) | 5.1\%) | 98.6\%) | 0.615 | 0.7137 |
| Score $\geq 3$ |  |  |  |  |  |  |
|  | 21.0\% | 91.8\% | 5.6\% | 98.0\% |  |  |
|  | (17.1\%- | (91.3\%- | (4.3\%- | (97.8\%- |  |  |
|  | 24.9\%) | 92.2\%) | 7.0\%) | 98.3\%) | 0.560 | <0.001 |
| Score $\geq 4$ |  |  |  |  |  |  |
|  | 8.9\% | 97.6\% | 8.0\% | 97.9\% |  |  |
|  | (6.2\%- | (97.4\%- | (5.1\%- | (97.6\%- |  |  |
|  | 11.6\%) | 97.9\%) | 10.9\%) | 98.1\%) | 0.533 | <0.001 |

*** P value compares AUC for a given score with the reference Score $\geq 1$

Table 6．Comparison of Receiver Operating Charac－ teristics Curve for HIV Risk Screening Tool （Score $\geq 1$ ）by Age，Gender，and Residence for Adults and Adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia （2016）and Tanzania（2016－2017）

|  | $\begin{aligned} & m \\ & 0 \\ & 0 \\ & 0 \end{aligned}$ |  |  | $\begin{aligned} & \infty \\ & \stackrel{\infty}{寸} \\ & 0 \\ & 0 \end{aligned}$ |  | $\stackrel{\bar{n}}{\stackrel{n}{*}}$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 |  |  |  |  |  |  |  |
|  | $\begin{aligned} & \bullet \\ & \stackrel{n}{n} \\ & n \\ & 0 \end{aligned}$ | $\begin{aligned} & n \\ & \frac{n}{n} \\ & 0 \end{aligned}$ | $\stackrel{\infty}{\stackrel{\infty}{*}}$ | $\begin{aligned} & \frac{2}{7} \\ & 0 \\ & \hline \end{aligned}$ | $\begin{aligned} & \underset{\sim}{N} \\ & \underset{0}{0} \end{aligned}$ | $\begin{aligned} & 0 \\ & \hat{8} \\ & 0 . \end{aligned}$ | $n$ 0 0 0 |
|  | $\begin{aligned} & \infty \\ & \infty \\ & \infty \\ & \underset{\sim}{\infty} \end{aligned}$ | $\begin{aligned} & \hat{o} \\ & \hat{n} \\ & n \end{aligned}$ | $\begin{aligned} & \stackrel{n}{f} \\ & \text { m } \end{aligned}$ | $\begin{aligned} & \text { n } \\ & \underset{\sim}{n} \end{aligned}$ | $\stackrel{n}{\infty}$ | ¢ | on <br>  <br> 0 |
|  | $\begin{gathered} \underset{\sim}{\lambda} \\ \stackrel{i}{n} \end{gathered}$ | $\begin{aligned} & \text { à } \\ & \underset{1}{n} \\ & \text { in } \end{aligned}$ | $\begin{aligned} & + \\ & i \end{aligned}$ | $\frac{0}{\pi}$ | $\begin{gathered} \frac{0}{\pi} \\ \text { E } \\ \text { U1 } \end{gathered}$ | 析 | 菏 |
|  | $\underset{8}{0}$ |  |  |  |  |  |  |

Figure 4 summarizes relationship between eligibility， sensitivity，and PPV or HIV testing yield．Eligibility for HIV test decreased with increasing risk score cut－ offs： $56 \%$ would be eligible with a cut－off score of $\geq 1$ ，while it was $2 \%$ for a cut－off score of $\geq 4$ ．HIV testing positivity（PPV）and sensitivity or eligibility was negatively correlated with both going down with increasing cut－off score，while PPV increased．


Figure 4．Relationship between Eligibility，Sensitivi－ ty，and HIV Testing Yield for a Risk Assessment Tool that contains all Risk Factors for Adults and Adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia （2016）and Tanzania（2016－2017）

In the tool that contained only statistically significant risk factors from the final logistic regression model （Table 3）（Tool 2），being DSW（odds ratio（OR）： 3.9 （ $95 \%$ CI：2．9－5．2）；being spouse of a known HIV＋ person（OR： 6.1 （ $95 \% \mathrm{CI}: 2.0-19.1$ ））；having history of selling sex for money（OR： 7.7 （ $95 \% \mathrm{CI}$ ：3．6－ 16．3））；having sexually transmitted infections in the past 12 months（OR： 2.1 （ $95 \% \mathrm{CI}: 1.4-3$ ）），and having a paid work in the past 12 months（OR： 2.1 （ $95 \%$ CI：1．4－3））；were included in the final model．Tool 4 which contained customized risk factors，the combi－ nation of risk factors having paid work in the past year and sleeping away from home for more than a month in the past 12 months were combined as pre－ dictors in addition to conventional risk factors．Hav－ ing a paid work and sleeping away from home were statistically significant predictors of undiagnosed HIV infection（OR： 1.8 （ $95 \% \mathrm{CI}: 1.1-3.0$ ））．

## Comparison of risk assessment tools

Looking at the different risk assessment tools，all were statistically significant predictors of HIV infec－ tion with p －value $<0.001$ ．For all tools，if none of the risk factors was present，HIV prevalence would be low in the range of $1.0-1.3 \%$ ．（Table 7）Sensitivity was better for Tool 1 but the corresponding specifici－ ty was the lowest．AUC was better for all other tools as compared to this tool，and the difference was much higher for Tools 3 and 4 （p－value $<0.001$ ）．（Table 8） PPV or HIV testing yield was highest for Tools 3 and 4 at $4.2 \%$ and $4.0 \%$ ，respectively，if at least one risk factor was present．Tool 3 had the lowest proportion of people eligible for testing at $34 \%$ the highest being for tool 1 at $59 \%$ ．（Table 7）Number needed to test （ $\mathrm{NNT}+$ ）was 24 for Tool 3，while it was 43 if univer－ sal testing was used．

Table 7．of Risk Scores with HIV Infection using a tool that contains all HIV Risk Factors for Adults and Adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia （2016）and Tanzania（2016－2017）

\#Only those included in the final model were considered; *Customized tool=Conventional risk factors + Working for a payment in the past 12 months and sleeping away from home for at least 1 month in the past 12 months of the survey.

Table 8. Sensitivity, Specificity, Positive Predictive Value (PPV+), and Negative Predictive Value (NPV) for each potential HIV Risk Screening Tool for Adults and Adolescents $>14$ years who were never tested for HIV before PHIA Surveys conducted in Zambia (2016) and Tanzania (2016-2017)

| Risk Factor Selection Strategy | Sensitivity | Specificity | PPV | NPV | $\underset{*}{\text { AUC* }}$ | $\begin{aligned} & \hline \mathbf{P} \\ & \text { value*** } \end{aligned}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Tool 1: All risk factors | $\begin{aligned} & \hline 82.3 \% \\ & (78.6 \%-85.9 \%) \end{aligned}$ | $\begin{aligned} & 41.9 \% \\ & (41.1 \%-42.7 \%) \end{aligned}$ | $\begin{aligned} & \text { 3.2\% } \\ & (2.8 \%-3.6 \%) \end{aligned}$ | $\begin{aligned} & 99.0 \% \\ & (98.8 \%-99.3 \%) \end{aligned}$ | 0.6116 |  |
| Tool 2: Statistically significant only ${ }^{\#}$ | $\begin{aligned} & 73.4 \% \\ & (69.1 \%-77.6 \%) \end{aligned}$ | $\begin{aligned} & 53.6 \% \\ & (52.8 \%-54.4 \%) \end{aligned}$ | $\begin{aligned} & 3.6 \% \\ & (3.1 \%-4.0 \%) \end{aligned}$ | $\begin{aligned} & 98.9 \% \\ & (98.6 \%-99.1 \%) \end{aligned}$ | 0.6267 | 0.035 |
| Tool 3: Conventional risk factors | $\begin{aligned} & 63.5 \% \\ & (58.9 \%-68.1 \%) \end{aligned}$ | $\begin{aligned} & 66.2 \% \\ & (65.5 \%-67.0 \%) \end{aligned}$ | $\begin{aligned} & 4.2 \% \\ & (3.6 \%-4.8 \%) \end{aligned}$ | $\begin{aligned} & 98.7 \% \\ & (98.5 \%-98.9 \%) \end{aligned}$ | 0.6469 | <0.001 |
| Tool 4: Customized tool* | $\begin{aligned} & \text { 65.5\% } \\ & \text { (61.0\%-70.1\%) } \end{aligned}$ | $\begin{aligned} & 63.4 \% \\ & (62.6 \%-64.2 \%) \end{aligned}$ | $\begin{aligned} & 4.0 \% \\ & (3.5 \%-4.5 \%) \end{aligned}$ | $\begin{aligned} & 98.7 \% \\ & (98.5 \%-99.0 \%) \end{aligned}$ | 0.6412 | $<0.001$ |

\#Only those included in the final model were considered; *Customized tool=Conventional risk factors + Working for a payment in the past 12 months and sleeping away from home for at least 1 month in the past 12 months of the survey. ${ }^{* *}$ AUC $=$ Area under the Curve of a Receiver Operating Curve;
*** P value compares AUC for a given risk assessment tool with the reference tool that contains all risk factors.

## Discussion

We set out to validate HIV risk assessment tool used for adults. In this process, we tried various combinations of risk factors in different tools for best possible
outcome. The final tool we recommend for use contains conventional risk factors. This screening tool showed a moderate sensitivity and specificity for identifying infected adults at household level. Using this screening tool, the number needed to test to diagnose one HIV infected adult would be 24 down from 43 if universal testing was used.
Looking at individual risk factors, the prevalence of HIV in those who never tested for HIV remained to be high compared to those without risk factors except for TB related risk factors and chronic illness. Two risk factors that stood out with having testing yield of $>10 \%$ were selling sex for money and having an HIV+ spouse. This is comparable to reported prevalence of $12-20 \%$ among FSWs in the study countries (23). ICT for spouses records even higher testing yield at $32 \%$ in program settings (24). Marital status is an important risk factor. Being divorced widowed or separated was found to be the third highest risk factor with a yield of $6.1 \%$. DSWs are easily identifiable at community level and can be used to identify at risk people at community or facility level. It is already a risk factor in many countries (25, 26).Cervical cancer is an important risk factor since Human Papiloma Virus, which is a sexually transmitted viral infection, is the causative agent (27). Coinfection with HIV was $5.5 \%$ in this study. Having multiple sexual partners was found to have a relatively lower prevalence at $3.1 \%$. This may be due to the higher condom use during casual sex with a nonregular partner $(28,29)$.
Lifetime TB disease was not statistically significant at the 0.05 cut-off. This should not be misinterpreted as TB not being a risk factor. Data on year of TB diagnosis were present only for Zambia and when we did analysis comparing TB diagnosed in the past 10 years to those who never had TB, or who had TB before 10 years, TB prevalence was much higher at $33.3 \%$ prevalence. This should be used in practice instead of lifetime TB disease.
Presumptive TB was not predictor of HIV infection because probably it was defined broadly especially for cough. A definition of cough $>2$ weeks may make improve the positivity. In studies where the later definition was used, the positivity was found to be higher ( 30,31 ).
Adults having multiple risk factors were found to have high testing yield and were a small fraction of the total assessed. This should be explored to further identify additional risk factors. A very good example in current use by different case finding and prevention programs is being long distance truck driver, who are likely to sleep away from home, and have multiple sexual partners including sex workers (32, 33).

This study also provides some form of reference for the percentage of people who are potentially eligible for HIV testing fulfilling at least one of the conventional risk factors among those who never tested for HIV. In this study, $34.4 \%$ adults who never tested for

HIV would be eligible for testing. That is around $9.1 \%$ of the initial number of adults interviewed. This provides a reference value with which to compare community HIV case finding interventions when such risk factors are used. However, it would not be advisable to test this much adults as it wouldn't be cost effective. A more targeted approach focusing on sex workers and their clients, partners of known HIV+ index cases, DSWs, and TB cases would be important starting points (2).
Eliciting some of the risk factors especially those related to sexual history may need some experience especially when implementing the risk assessment tool at community level. The use of health extension workers or community health workers who formally do health interventions may help. At facility level where these risk factors are often used maintaining quality of counselling needs to be ensured through ongoing training and on job coaching. Some of the risk factors are treated in speciality clinics like TB in TB clinic, or STI and cervical cancer in gynaecology clinics for women. This will make it easier to implement universal testing for these groups by providing integrated testing services.
The large number of study participants was one of the strengths of the study. Missing data was minimal and was not related to the risk factors being studied. The performance of the final tool was found to be independent of age, gender, and residence making the use of the tool applicable in different scenarios. Some of the risk factors were captured a little different from what is used in actual settings. All parameters of screening tool are likely to improve if the presence of the following risk factors was determined for the past 10 years just like what we did with TB instead of just the past 12 months: multiple sexual partners, STI, and clients of sex workers.

## Conclusion

Use of a screening tool containing conventional risk factors improved HIV testing yield compared to doing universal testing. The use of multiple risk factors to improve HIV testing yield should be explored further.

## Declarations

## Consent for publication

Not applicable

## Conflict of interest

The authors declare that they have no competing interests.

## Author contribution

KDY originated the research idea, and collected and analyzed the data; KDY \& JM contributed to data analysis and writing the manuscript; All authors read and approved the final manuscript.

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[^0]:    ${ }^{5}$ Divorced, Separated, or Widowed; *within the last 12 months of the survey, ${ }^{\text {\# }}$ Cough, fever, night sweats, or

