# Establishment of a predictor risk score for the prioritization of patients for testing for acute HIV infection



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## Background

- ➤ Resource-poor settings rarely screen for acute HIV infection (AHI).
- ➤ AHI exists during the time period between HIV infection and HIV sero-conversion.
- > Not addressing AHI may jeopardize HIV epidemic control.
- ➤ One barrier is the lack of contextualized screening algorithms that would allow prioritization of patients for more resource-intensive diagnostic viral load (VL) testing.

## **Objective**

To develop a predictor risk score (PRS) algorithm that may assist health workers to select patients for AHI testing in Eswatini.

## Methods

- ➤ Adult outpatients with a HIV-negative or discordant test result using serial Alere™ Determine and Uni-Gold™ testing algorithms underwent VL testing (Xpert®) for the diagnosis of AHI at Nhlangano Health Centre, from March 2019 to March 2020.
- ➤ AHI definition: VL ≥40 copies/mL and HIV sero-negative/ discordant RDT.
- A nurse performed physical examination and administered questionnaires assessing AHI risk factors.

## **Statistics**

- We used the least absolute shrinkage and selection operator (Lasso) method to determine factors for AHI prediction.
- Their beta-coefficients were rounded to the nearest integer to obtain predictor scores for each patient.
- Test characteristics of the PRS of the entire cohort for identification of AHI were described in comparison with Xpert testing results.
- Finally, the performance of four external PRS reported from Africa was assessed with receiver operating characteristic (ROC) curve statistics.

## **Ethical approval**

Ethical approval was obtained from the Eswatini Health and Human Research Review Board, and the MSF Ethics Review Board.

## Acknowledgements

Patients and health workers of Nhlangano Health Centre; Ministry of Health: ENAP, NRL; Current and former MSF staff.

## RESULTS

#### **Study enrolment**

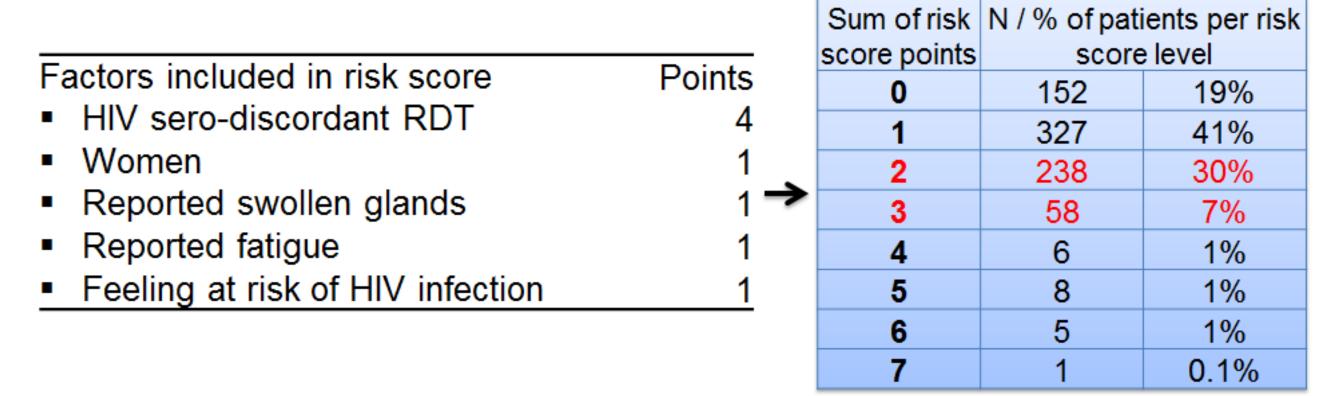
➤ Of 795 patients tested, 30 (3.8%) presented with AHI.

#### Predictor risk score (PRS)

- The final PRS comprised the following factors with rounded beta-coefficients/risk scores: discordant rapid-diagnostic test result (4), female sex (1), feeling at risk of HIV (1), self-reported swollen glands (1), and fatigue (1); see Figure 1.
- As identified with ROC statistics, the PRS performed best for patients with a risk-score cutoff at ≥2 or ≥3 points (Figure 2).

# Performance of PRS at two cut-off points (Table 1)

- At the cut-off of ≥2 points, sensitivity and specificity were 86.7% and 62.1%.
- ➤ At the cut-off of ≥3 points, sensitivity decreased to 50.0% and specificity increased to 91.8%.
- While NPV was ≥97.9% for both cutoff points, the PPV remained at ≤19.2%.



<u>Figure 1:</u> Factors identified in for inclusion into the PRS, and number and proportion of patients falling into the different risk score categories.

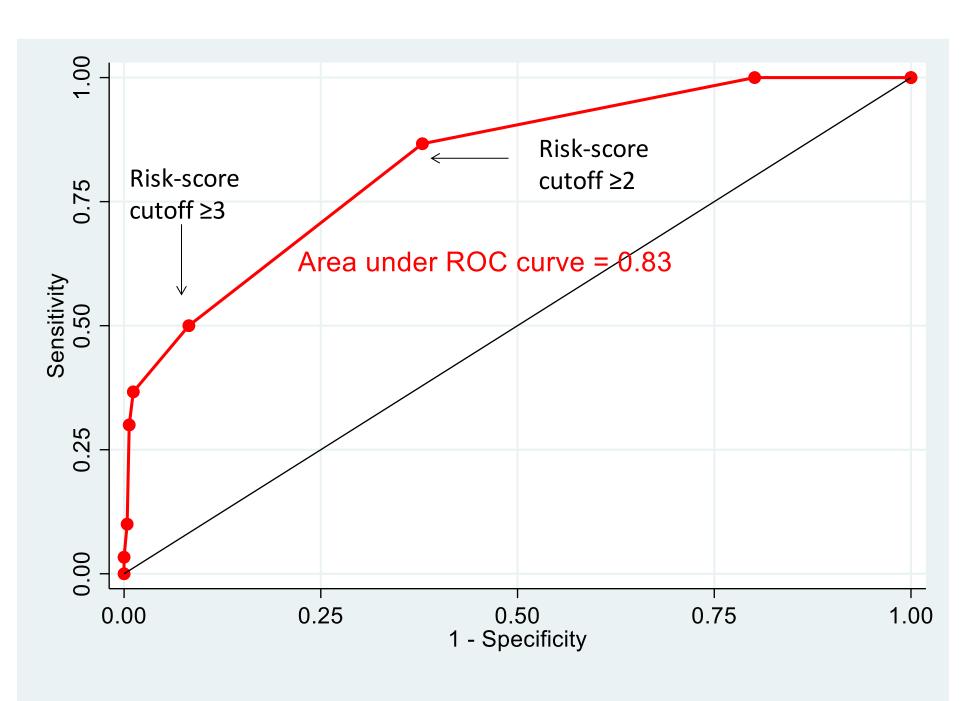


Figure 2: The best risk score cut-off levels were identified with ROC statistics.

The cut-off ≥2 and ≥3 points were identified with most favorable sensitivity and specificity estimates for prediction of AHI.

<u>Table 1:</u> Performance indicators of the PRS at two cut-off levels for the identification of AHI patients. Point estimate and 95% CI presented.

|                           | PRS cutoff ≥2 (n=795) |                 | PRS cutoff ≥3 (n=795) |                 |
|---------------------------|-----------------------|-----------------|-----------------------|-----------------|
| Sensitivity               | 86.7%                 | (69.3% - 96.2%) | 50.0%                 | (31.3% - 68.7%) |
| Specificity               | 62.1%                 | (58.5% - 65.5%) | 91.8%                 | (89.6% - 93.6%) |
| ROC area                  | 0.74                  | (0.68 - 0.81)   | 0.71                  | (0.62 - 0.80)   |
| Positive predictive value | 8.2%                  | (5.5% - 11.8%)  | 19.2%                 | (11.2% - 29.7%) |
| Negative predictive value | 99.2%                 | (97.9% - 99.8%) | 97.9%                 | (96.6% - 98.8%) |

## **Comparison with other external PRS**

➤ Based on ROC statistics, the study-specific PRS (ROC 0.83) had the highest ability to correctly classify AHI cases while ROC statistics for external PRS ranged from 0.50 to 0.74.

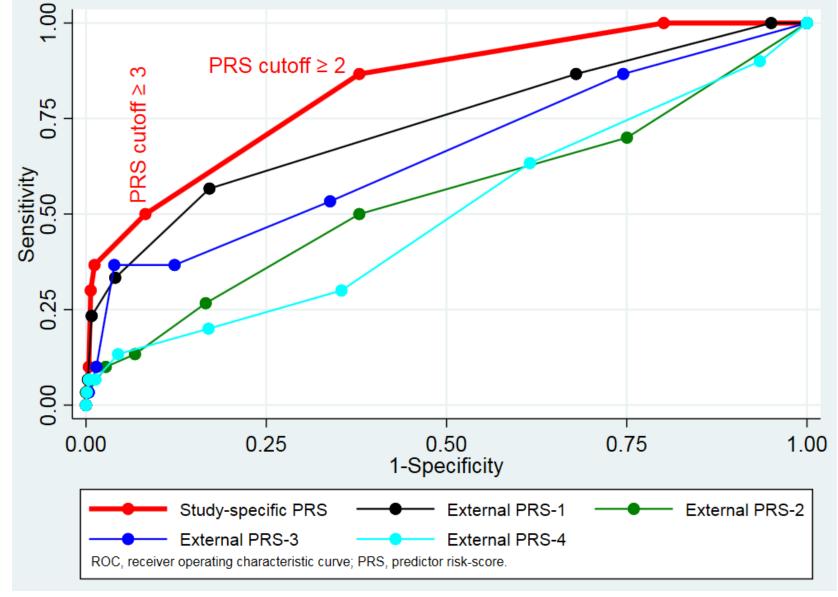


Figure 3:
Performance of
the studyspecific PRS
compared with
PRS from other
settings in SubSaharan Africa.

## Conclusions

- > PRS can identify patients at risk of AHI, enabling prioritization for diagnostic viral load testing.
- Further studies should evaluate the routine use of PRS in public sector settings and validate external PRS before local use.

