

# Response to reviewers and editorial comments for “Spatio-temporal estimates of HIV risk group proportions for adolescent girls and young women across 13 priority countries in sub-Saharan Africa”

Corresponding author: Adam Howes ([ath19@ic.ac.uk](mailto:ath19@ic.ac.uk))

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# Reviewer 1

We thank the reviewer for their helpful comments regarding the statistical modelling.

*1. In the manuscript, explain why you use the INLA not WINBUGS coding? The multinomial regression could be modeled in WINBUGS directly.*

The reviewer is right to note that multinomial logistic regression models can be implemented in probabilistic programming languages like `WinBUGS` directly. However, for this application, Markov chain Monte Carlo approaches would be prohibitively expensive. For this reason, we chose to use integrated nested Laplace approximations via `R-INLA`, which have been shown to have comparable accuracy for latent Gaussian models in the realistic, pre-asymptotic regime. We have added the following text to the methods section of the manuscript to clarify this point:

For models with a Gaussian latent field, INLA has comparable accuracy to Markov chain Monte Carlo in the realistic, pre-asymptotic regime, and is substantially more computationally tractable for high dimensional models like ours, which has 940 districts, 20 years, 3 age groups, and 4 risk groups.

*2. Is there no other potential covariate that could be used for better modeling?*

Many of the covariates which one might expect to be most predictive of risk group proportions are themselves difficult to accurately measure, and can only lead to modest at best improvements in model performance. For example, despite the case for their being a clear link between the “proportion clients of FSW” covariate and the “proportion FSW” outcome we found only marginal benefits to inclusion.

We agree that identifying predictive and measureable predictors of risk group proportions, or high risk locations, is an important area for further research. We have commented on this in the discussion section:

More generally, we did not aim to include predictive covariates. Many covariates that might be predictive of risk group proportions (for example sociodemographic characteristics, education, local economic activity, cultural and religious norms and attitudes) are themselves difficult to measure spatially, and previous efforts have only led to modest improvements in model performance. Identifying measurable correlates of risk, or particular settings in which time-concentrated HIV risk occurs, is an important area for further research to improve risk prioritisation and precision HIV programme delivery.

*3. Please explain sub-national effect more clearly. Why and how you used it?*

We did not use the specific term “sub-national effect” in the manuscript. We believe the reviewer is referring to the term “spatial effect”, which we use to describe district-specific effects that allow the model to capture district-level variation in risk group proportions, primarily informed by survey data in each district.

Thank you to the reviewer for highlighting the omission of specific definition for this term. Throughout the manuscript, we have replaced references to “effect” by the more accurate “random effect”, and have clarified the usage of spatial random effect where it first appears in the methods section as follows:

All models included intercepts for each risk group, as well as age, country, and age-country random effects. To account for district-level variation we used spatial random effects consisting of a parameter for each district. We considered alternative model specifications in which the spatial random effects were either independent or spatially correlated such that more information was shared between neighboring districts than those far apart. Similarly, we used temporal random effects to allow variation in risk group proportions over time, and considered alternative model specifications as independent versus first-order auto-regressive where a smooth temporal trend is assumed.

*4. Why the interaction term for spatiotemporal effect didn't consider in the modeling framework?*

We share the reviewer’s interest in the spatiotemporal interaction in risk group proportions, which would allow variation in temporal trend by country. However, we found that these interactions made the model computationally intractable to estimate given the large number of districts, years, age groups and risk groups

involved. This is a limitation of our analysis, but, in practice, we do not believe this has a large impact on our results because overall we found that risk group proportions were very stable over time, while they varied substantially spatially. More specifically, when we fit the model to each country individually, which can be considered as an extreme version of country-specific spatiotemporal interactions, the proportion of variance (Sobol' index) attributable to the temporal (survey) random effects<sup>1</sup> was on average 2.5800954/% (Supplementary Figure B.3). This is corroborated by the lack of temporal trends in Supplementary Figures B.5 through B.17 which show the modelled and direct estimates for each country individually, as well as the fact that unstructured (IID) rather than structured (AR1) temporal random effects were preferred in the model selection (Supplementary Figure A.1).

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<sup>1</sup>For those countries where such a random effect could be meaningfully defined.

## Reviewer 2

*This is a well-crafted manuscript investigating the spatio-temporal estimates of HIV risk group proportions for adolescent girls and young women across 13 priority countries in sub-Saharan Africa. Their analyses identify specific age groups at the district level that should be targeted for HIV intervention in SSA. This is critical in reducing the HIV epidemic in the southern region of SSA. In addition to the main figures, the supplementary Tables and Figures show country-by-country risk, mostly among female sex workers for all age groups. With the help of their models, specific resources can target at-risk populations with a moderate assurance of how many people to reach and where these resources should go.*

We thank the reviewer for their kind comments.

*My little concern is about using different data from UNAIDS Key Population Atlas apart from the DHS, which is the main source data for the analyses. I believe the two variants of data are based on different designs, and combining them may not result in dependable results. It would have been more attainable if the UNAIDS data had been used in their sensitivity analysis to confirm the results from the DHS data.*

For the FSW risk group, we used age-disaggregated (Supplementary Figure A.4) national-level estimates from Stevens et al. (2022) to inform the national risk group population size within each age group, and household survey data to inform subnational variation. Estimates of hidden populations like FSW from household surveys have significant limitations due in part to stigma around disclosing membership, as well as potential for not being included in the sampling frame (Abdul-Quader, Baughman, and Hladik 2014). For this reason, we believe it is more appropriate to calibrate our estimates to Stevens et al. (2022), who as well as including the KP Atlas data, integrate data from other FSW population size estimates using a Bayesian mixed effects model.

Overall, we agree with the reviewer’s comment that relying on data from different designs is not the ideal analysis, but we believe that it is the best approach using available data given known limitations identifying women who sell sex in household sampling frames and questionnaires. We have noted this in the limitations section of the discussion:

We have the least confidence in our estimates for the FSW risk group. As well as having the smallest sample sizes, our transactional sex estimates do not overcome the difficulties of sampling hard to reach groups. We inherent any limitations of the national FSW estimates which we adjust our estimates of transactional sex to match. Furthermore, we do not consider seasonal migration patterns, which may particularly affect FSW size.

*Figure 1 is not clear. I recommend that the authors use a table as an alternative visualization.*

In our and other previous work, figures similar to Figure 1 have been effective at concisely visualising inclusion of a large number of surveys across several countries and types (see, for example, Figure 1 of Giguère et al. (2021)), and we prefer to retain this figure. Supplementary Table B.3 provides an alternative tabular summary of the surveys analysed, with sample size broken down by age group. We have extended the caption of Figure 1 to draw attention to this alternative presentation:

**Fig 1.** Surveys used in our analysis by year, survey type, sample size, and whether the survey included a question about transactional sex. Details of included surveys are in Supplementary Table B.3.

## Editorial comments

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*Figure 2: please (a) provide a direct link to the base layer of the map (i.e., the country or region border shape) and ensure this is also included in the figure legend; and (b) provide a link to the terms of use / license information for the base layer image or shapefile. We cannot publish proprietary or copyrighted maps (e.g. Google Maps, Mapquest) and the terms of use for your map base layer must be compatible with our CC-BY 4.0 license. Note: if you created the map in a software program like R or ArcGIS, please locate and indicate the source of the basemap shapefile onto which data has been plotted. If your map was obtained from a copyrighted source please amend the figure so that the base map used is from an openly available source. Alternatively, please provide explicit written permission from the copyright holder granting you the right to publish the material under our CC-BY 4.0 license. Please note that the following CC BY licenses are compatible with PLOS license: CC BY 4.0, CC BY 2.0 and CC BY 3.0, meanwhile such licenses as CC BY-ND 3.0 and others are not compatible due to additional restrictions. If you are unsure whether you can use a map or not, please do reach out and we will be able to help you. The following websites are good examples of where you can source open access or public domain maps: U.S. Geological Survey (USGS) - All maps are in the public domain. (<http://www.usgs.gov>), PlaniGlobe - All maps are published under a Creative Commons license so please cite "PlaniGlobe, <http://www.planiglobe.com>, CC BY 2.0" in the image credit after the caption. (<http://www.planiglobe.com/?lang=enl>), Natural Earth - All maps are public domain. (<http://www.naturalearthdata.com/about/terms-of-use/>)*

Response here.

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