

## RESEARCH ARTICLE

# Estimating the contribution of HIV-infected adults to household pneumococcal transmission in South Africa, 2016–2018: A hidden Markov modelling study

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**Data Availability Statement:** Data cannot be shared publicly because of confidentiality. Data are available from the South Africa National Institute of Communicable Disease (NICD) if authorised by the Ethics Committee for researchers who meet the criteria for access to confidential data. Please contact The Chair person Human Research Ethics Committee University of Witwatersrand Phillip V Tobia Building CNR York Road & 29 Princess of Wales Terrace Johannesburg South Africa Email: [Rhulani.Mkansi@wits.ac.za](mailto:Rhulani.Mkansi@wits.ac.za) Email: [zanele](mailto:zanele).

## Abstract

Human immunodeficiency virus (HIV) infected adults are at a higher risk of pneumococcal colonisation and disease, even while receiving antiretroviral therapy (ART). To help evaluate potential indirect effects of vaccination of HIV-infected adults, we assessed whether HIV-infected adults disproportionately contribute to household transmission of pneumococci. We constructed a hidden Markov model to capture the dynamics of pneumococcal carriage acquisition and clearance observed during a longitudinal household-based nasopharyngeal swabbing study, while accounting for sample misclassifications. Households were followed-up twice weekly for approximately 10 months each year during a three-year study period for nasopharyngeal carriage detection via real-time PCR. We estimated the effect of participant's age, HIV status, presence of a HIV-infected adult within the household and other covariates on pneumococcal acquisition and clearance probabilities. Of 1,684 individuals enrolled, 279 (16.6%) were younger children (<5 years-old) of whom 4 (1.5%) were HIV-infected and 726 (43.1%) were adults ( $\geq 18$  years-old) of whom 214 (30.4%) were HIV-infected, most (173, 81.2%) with high CD4+ count. The observed range of pneumococcal carriage prevalence across visits was substantially higher in younger children (56.9–80.5%) than older children (5–17 years-old) (31.7–50.0%) or adults (11.5–23.5%). We estimate that 14.4% (95% Confidence Interval [CI]: 13.7–15.0) of pneumococcal-negative

ndlovu@wits.ac.za URL: <https://www.wits.ac.za/research/researcher-support/research-ethics/ethics-committees/> The code underlying the results presented in the study are available from GitHub through the following link (<https://github.com/deusthindwa/hmm.pneumococcus.hiv.south-africa>).

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swabs were false negatives. Daily carriage acquisition probabilities among HIV-uninfected younger children were similar in households with and without HIV-infected adults (hazard ratio: 0.95, 95%CI: 0.91–1.01). Longer average carriage duration (11.4 days, 95%CI: 10.2–12.8 vs 6.0 days, 95%CI: 5.6–6.3) and higher median carriage density (622 genome equivalents per millilitre, 95%CI: 507–714 vs 389, 95%CI: 311.1–435.5) were estimated in HIV-infected vs HIV-uninfected adults. The use of ART and antibiotics substantially reduced carriage duration in all age groups, and acquisition rates increased with household size. Although South African HIV-infected adults on ART have longer carriage duration and density than their HIV-uninfected counterparts, they show similar patterns of pneumococcal acquisition and onward transmission.

## Author summary

We assessed the contribution of HIV-infected adults to household pneumococcal transmission by applying a hidden Markov model to pneumococcal cohort data comprising 115,595 nasopharyngeal samples from 1,684 individuals in rural and urban settings in South Africa. We estimated 14.4% of sample misclassifications (false negatives), representing 85.6% sensitivity of a test that was used to detect pneumococcus. Pneumococcal carriage prevalence and acquisition rates, and average duration were usually higher in younger or older children than adults. The use of ART and antibiotics reduced the average carriage duration across all age and HIV groups, and carriage acquisition risks increased in larger household sizes. Despite the longer average carriage duration and higher median carriage density in HIV-infected than HIV-uninfected adults, we found similar carriage acquisition and onward transmission risks in the dual groups. These findings suggest that vaccinating HIV-infected adults on ART with PCV would reduce their risk for pneumococcal disease but may add little to the indirect protection against carriage of the rest of the population.

## Introduction

*Streptococcus pneumoniae* (pneumococcus) caused an estimated 3.7 million cases of invasive pneumococcal disease (IPD) and 317,300 deaths in children <5 years-old, globally in 2015 [1,2]. While severe disease is largely concentrated in young children and older adults, human immunodeficiency virus (HIV)-infected adults are also at an increased risk of both colonisation and IPD [3–7]. HIV affects the T and B cell function, resulting in impaired responses to control pneumococcal carriage at mucosal level [8–10]. Although the universal scale-up of antiretroviral therapy (ART) [11,12] has successfully reduced IPD risk in HIV-infected adults [13,14], the IPD risk remains elevated if compared to HIV-uninfected adults [5,6]. ART partially reconstitutes mucosal immunity by increasing B and T cell quantity and functionality [8,15], but deficiencies in humoral mucosal response due to depleted or persistent defects in memory cell function persist after ART initiation [16–18].

Despite mature pneumococcal conjugate vaccine (PCV) infant immunisation programmes, continued circulation of vaccine preventable serotypes in adults has been observed throughout Africa [19–25]. In some countries, such as Malawi, Mozambique, and Kenya, this intersects with areas of high HIV prevalence. Adult HIV prevalence in Africa remains high [26] as a consequence of improved survival with ART use [11,12] and persistently high HIV incidence [27],

thus the high risk of pneumococcal carriage and IPD in HIV-infected adults in Africa remains a concern.

Presently, no pneumococcal immunisation program for HIV-infected adults exist in South Africa and low-income African countries [28]. Vaccination of African HIV-infected adults with PCV, similar to the recommendations in many high-income countries, may not only reduce their disease burden but also vaccine serotype pneumococcal acquisition and hence onward transmission and may thus benefit non-vaccinated populations [29]. We hypothesised that children living with HIV-infected adults have higher rates of pneumococcal carriage acquisition due to increased exposure from frequently colonised HIV-infected adults who usually have a prolonged higher carriage prevalence [5]. In this study, we assessed whether HIV-infected adults contribute more to pneumococcal transmission within the household than their HIV-uninfected counterparts.

## Methods

### Ethics statement

The longitudinal pneumococcal carriage data described in this study were obtained from South African children and adults through a written consent as part of the PHIRST study. For a child participant, written consent was obtained from a parent or guardian. The use of data was granted by the University of Witwatersrand, Human Research Ethics Committee (HREC) and the Protocol Review Committee (PRC) under approval #150808, the US CDC's Institutional Review Board relied on the local review (#6840), and the London School of Hygiene & Tropical Medicine Observational Research Ethics Committee under approval #17902.

### Data description

The temporal dynamics of pneumococcal colonisation were observed in a cohort study (Prospective Household Observational Cohort Study of Influenza, Respiratory Syncytial Virus and Other Respiratory Pathogens Community Burden and Transmission Dynamics—PHIRST) conducted between 2016 and 2018 in a rural (Agincourt) and an urban (Klerksdorp) community in South Africa. Households were randomly selected, and were eligible for the study if they had  $\geq 3$  household members and the household members resided in the household for  $\geq 1$  year prior to study commencement, had no plan to relocate during study duration, and consented to participate in the study. Also, enrolment ensured that more than half of the households included at least one child aged  $< 5$  years, and a new cohort was enrolled every study year [30,31].

A total of 1,684 individuals from 327 households were enrolled and followed up from May to October in 2016 and January to October in 2017 and 2018. The median household size was 5 (interquartile range 4–7). Nasopharyngeal (NP) swabs were taken twice weekly, resulting in 115,595 total NP samples from 1,684 individuals. The swabs were tested for the presence of pneumococci using real-time quantitative polymerase chain reaction (qPCR), targeting the autolysin (*lytA*) gene [32]. Serotyping was not performed. On enrolment, the demographic characteristics of the study participants were recorded, and household members were tested for HIV infection according to the double rapid test algorithm in South Africa [33]. Participants were considered HIV infected if they had two positive rapid HIV tests, evidence of a positive HIV laboratory result or evidence of ART treatment. Participants were considered HIV uninfected if they had a documented negative HIV test result. A documented HIV negative status for the mother confirmed HIV negative status for a child aged  $< 10$  years. HIV infection was confirmed by PCR in children aged  $< 18$  months. In all HIV infected individuals, specimens were collected for CD4+ T cell and quantitative HIV viral load testing. Newly HIV diagnosed patients were referred to the local HIV/ART clinic [30].

## Modelling framework

We used a continuous time, time homogeneous, hidden Markov model (HMM) which assumed a Susceptible–Infected–Susceptible (SIS) framework [34–40], to fit to individual level trajectories of colonisation during the study period. An individual can be either infected (I or 2) and currently carrying pneumococci or be susceptible (S or 1). Thus, the model can be described by transition intensities between S and I for acquisition ( $q_{12}$ ) and clearance ( $q_{21}$ ) in the transition intensity matrix  $Q = \begin{pmatrix} -q_{12} & q_{12} \\ q_{21} & -q_{21} \end{pmatrix}$ . The effect ( $\beta$ ) of  $i^{\text{th}}$  individual covariate ( $z_i$ ) for acquisition and clearance rates over all transitions ( $T$ ) are incorporated via proportional hazard models  $q_{12}(z_i(t)) = q_{12}^{(0)} \exp(\beta_{12}^T z_i(t))$  and  $q_{21}(z_i(t)) = q_{21}^{(0)} \exp(\beta_{21}^T z_i(t))$ , respectively. To obtain the transition probabilities, matrix  $P = \begin{pmatrix} 1 - p_{12} & p_{12} \\ p_{21} & 1 - p_{21} \end{pmatrix}$  is defined and explicitly calculated through matrix exponential,  $P = \exp(Q(t))$ , where  $p_{12}$  is the probability of being in state 2 (I) at time  $t > 0$ , given that the previous state was 1 (S). A more detailed description of the Markov transition process is provided in the Supplement.

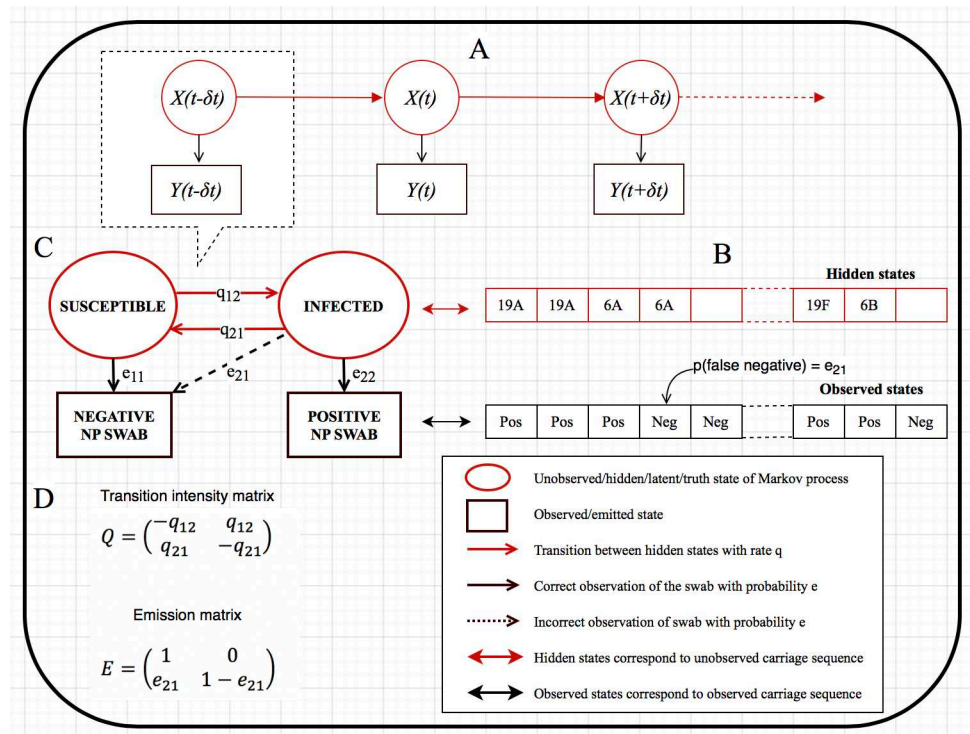
In the hidden Markov modelling (HMM) framework [36,41–46], the states S and I of the Markov Chain ( $X_i(t)$ ) for individual  $i$  at time  $t$  are not observed directly, but approximated by the results of a NP swab. The link between the modelled, true infection status and observed pneumococcal carriage states in the model ( $Y_i(t)$ ) is governed by emission probabilities conditional on the unobserved state. We assumed 100% specificity of the NP swab and the PCR (no false positive) while estimating the proportion of false negative results ( $e$ ) probabilistically (observed vs hidden/truth states). Hence, the emission matrix is given as  $E = \begin{pmatrix} 1 & 0 \\ e_{21} & 1 - e_{21} \end{pmatrix}$  where  $e_{21} = \Pr(Y_i(t) = 1 | X_i(t) = 2)$ .

We assumed that the observed states are conditionally independent given the values of the unobserved states and that the future Markov chain is independent of its history beyond the current state (Markov property) (Fig 1). Thus, the likelihood is the product of the emission probability density and the transition probability of hidden Markov chain summed over all possible paths of the hidden states (explicitly defined in the Supplement).

Our model assumed that carriage acquisition at the current observation point was a function of individual age group (younger child aged <5 years, older child aged 5–17 years, or adult aged  $\geq 18$  years), HIV status (infected or uninfected), number of HIV-infected adult(s) in the household, place of carriage exposure (household or community), and household size. Carriage duration was modified by individual age, HIV status, ART status, and antibiotic use. The place of carriage exposure is generally unknown without fine-scale serotype data. Crudely, we assumed that if a household member is currently infected while all other household members were susceptible at the last observation point, then current carriage acquisition of that member was attributable to community transmission [34]. Otherwise, we assumed that the transmission was from within the same household (Fig 1).

## Model fit, convergence and prediction

The model was fitted to longitudinal data of pneumococcal carriage dynamics in a maximum likelihood framework using Bound Optimisation By Quadratic Approximation optim algorithm facilitated by msm R package [36,47]. To ascertain convergence of the model, we purposefully selected five unique pairs of initial transition intensities {S, I} for the Q matrix, then refitted the model five times, each time starting a Markov chain with a unique dyad and



**Fig 1. Susceptible-infected-susceptible (SIS) hidden Markov model schemas of pneumococcal carriage dynamics in South African households between 2016–2018.** Continuous-time time-homogeneous hidden Markov model where  $X(t)$  represents hidden states, and  $Y(t)$  observed states, and in which  $Y(t)$  is conditionally independent given  $X(t)$  and the Markov property holds (A). Pneumococcal nasopharynx (NP) carriage sequence of a specified individual representing hidden and observed states, with a probability that an individual truly carrying a pneumococcal serotype may be detected negative by a real-time quantitative polymerase chain reaction test (B). An SIS hidden Markov model structure that captures a snapshot of part A and carriage sequence of part B in order to estimate transition rates and probability of misclassification/false negative (C). Transition intensity matrix,  $Q$ , and emission matrix,  $E$ , respectively capture the SIS model transition rates and emission or misclassification probability in part C to compute the maximum likelihood estimates of transition intensities and misclassification probability (D).

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iterating 1,000 times to obtain similar final transition intensities and  $-2\log$ -likelihood. Model predictions were assessed by comparing infection and susceptibility prevalence in 14-day intervals for the observed data to the fitted values. (S1 Fig in S1 File) [36].

### Decoding the underlying carriage sequence

After fitting the HMM, a Viterbi algorithm with the `msm` function was used to recursively construct the sequence of pneumococcal carriage with the highest probability through the hidden states [48]. The probability of each hidden state at each observation point, conditionally on all the data was computed using Baum-Welch forward/backward algorithm. Thus, an overall misclassification probability of the observed states given the hidden states was computed. Model estimates of carriage transition intensity and probability were adjusted for misclassification probability (S2 Fig in S1 File).

### Sensitivity analysis

In a sensitivity analysis, three alternative and potentially more parsimonious models were fitted separately to the data. Fits of these models were compared to the main model using Akaike Information Criterion (AIC) [49] and checked whether they yielded qualitatively different

results to the main model. Each of the four fitted models assumed the same number of covariates to modify carriage acquisition intensity but varying number of covariates assumed to modify carriage duration. Potential modifiers of carriage duration included age and HIV status for model 1; age, HIV status and antibiotic use for model 2; age, HIV status and viral load based ART status for model 3; and age, HIV status, antibiotic use and viral load based ART status for main model 4 (S1 Table).

Further, we examined the impact of alternative stratification of covariates on the changes in carriage transition probabilities: (i) while the main analysis estimated age- and HIV-stratified carriage acquisition rates comparing households with  $\geq 1$  HIV-infected adult(s) versus households without HIV-infected adults, in the sensitivity analysis, we estimated age- and HIV-stratified carriage acquisition rates comparing households with 0, 1, 2, 3, 4 and 5 HIV-infected adult(s) and (ii) rather than assuming time-homogeneous intensities throughout the study period, we relaxed this assumption by fitting a time-inhomogeneous model with yearly piecewise follow-up periods; 2016, 2017, and 2018 (S3 Fig in S1 File).

Statistical significance was set at  $< 0.05$ . All analyses were conducted in R v3.5.0 [36,50] and are available via <https://github.com/deusthindwa/hmm.pneumococcus.hiv.south-africa>.

## Results

### Descriptive analysis

A total of 327 households were recruited in the PHIRST study of which 166 (50.8%) had at least one member living with HIV infection. At enrolment, of 1,684 individuals included in the study, 279 (16.6%) were younger children aged less than 5 years old of whom 4 (1.5%) were HIV-infected, and 679 (40.3%) were older children aged between 5–17 years old of whom 31 (4.7%) were HIV-infected. Among the 726 (43.1%) study participants aged 18 years or older (“adults”), 214 (30.4%) were HIV-infected, and 505 (69.6%) were females. Among those HIV-infected adults, 196 (86.7%) self-reported to be on ART, although only 151 (79.5%) had CD4+ cell count of more than 350. Most adults were non-smokers (69.6%) and did not regularly consume alcohol (57.4%). A similar proportion of children lived in households with (50.6%) and without (49.4%) at least one HIV-infected adult. Among 231 younger children with vaccine information available, 227 (98.3%) received first PCV dose at 6 weeks, 225 (97.4%) second dose at 14 weeks and 216 (93.5%) third dose at 9 months of age (Table 1).

### Carriage prevalence and density

We estimated carriage prevalence by dividing the number of PCR positive samples by the number of swabs taken per visit per age or HIV group. Among HIV-uninfected participants, observed pneumococcal carriage prevalence was higher in younger children (range across visits: 56.9–80.5%,  $n = 256$ ) than older children (31.7–50.0%,  $n = 634$ ) and was lowest in adults (11.5–23.5%,  $n = 489$ ) (Fig 2A). Among HIV-infected participants, pneumococcal carriage prevalence fluctuated in younger children (0–100%,  $n = 4$ ), in older children (30–77%,  $n = 31$ ), and in adults (14–34%,  $n = 214$ ) (Fig 2A). The likelihood of detecting pneumococcal carriage during visits was higher for children than adults and for HIV-infected younger children or older children or adults than their HIV-uninfected counterparts (Fig 2B). Carriage prevalence among younger HIV-uninfected children was lower in households with less than 6 members (65.5%, 95%CI: 64.5–66.5) than in households with 6–10 (72.5%, 95%CI: 71.5–73.5) or household more than 10 members (85.6%, 95%CI: 82.4–88.8) but it was similar in HIV-infected children across household size groups (Fig 2C). Carriage prevalence fluctuated across visits by HIV-infection and sex in adults, with similar ranges between HIV-uninfected male adults

**Table 1. Baseline demographic and clinical characteristics of children and adults who were followed up twice weekly for ten months for nasopharynx swabbing for pneumococcal carriage in South African households between 2016 and 2018.**

Description	Total	Younger children	Older children	Adults
		<5 years	5–17 years	≥18 years
	<b>N = 1,684</b>	<b>n = 279 (16.6%)</b>	<b>n = 679 (40.3%)</b>	<b>n = 726 (43.1%)</b>
Mean age (SD)	22.1 (±19.8)	2.2 (±1.3)	10.5 (±3.7)	40.5 (±16.7)
Study site				
Agincourt (rural)	849 (50.4)	171 (61.3)	376 (55.4)	302 (41.6)
Klerksdorp (semi-urban)	835 (49.6)	108 (38.7)	303 (44.6)	424 (58.4)
Sex				
Female	1,009 (59.9)	137 (49.1)	367 (54.1)	505 (69.6)
Male	675 (40.1)	142 (50.9)	312 (45.9)	221 (30.4)
HIV status				
Negative (-)	1,379 (84.7)	256 (98.5)	634 (95.3)	489 (69.6)
Positive (+)	249 (15.3)	4 (1.5)	31 (4.7)	214 (30.4)
Viral-load based ART status <sup>†</sup>				
Not on ART	93 (40.8)	1 (50.0)	13 (52.0)	79 (39.3)
On ART	135 (59.2)	1 (50.0)	12 (48.0)	122 (60.7)
Self-reported ART status				
Not on ART	30 (13.2)	0 (0.0)	1 (3.8)	29 (14.5)
On ART	198 (86.8)	2 (100.0)	25 (96.2)	171 (85.5)
Mean CD4 count (SD) CD4+ cell count <sup>‡</sup>	673 (±430)	1,210 (±13.4)	857 (±418)	645 (±426)
Low	49 (22.8)	0 (0.0)	10 (43.5)	39 (20.5)
High	166 (77.2)	2 (100.0)	13 (56.5)	151 (79.5)
Living with ≥1 HIV+ adults				
No	818 (48.7)	133 (47.8)	346 (51.1)	339 (46.9)
Yes	860 (51.3)	145 (52.2)	331 (48.9)	384 (53.1)
PCV13 doses received <sup>#</sup>				
At 6 weeks	227 (98.3)	227 (98.3)	N/A	N/A
At 14 weeks	225 (97.4)	225 (97.4)	N/A	N/A
At 9 months	216 (93.5)	216 (93.5)	N/A	N/A
Smoking (≥18 years)				
No	505 (69.6)	N/A	N/A	505 (69.6)
Yes	221 (30.4)	N/A	N/A	221 (30.4)
Alcohol use (≥18 years)				
No	417 (57.4)	N/A	N/A	417 (57.4)
Yes	309 (42.6)	N/A	N/A	309 (42.6)

<sup>†</sup> ART use status based on viral load results (on ART = Undetected; Not on ART = <20 copies per ml)

<sup>‡</sup> Low CD4+ count ≤350 cells/mm<sup>3</sup> and high CD4+ count >350 cells/mm<sup>3</sup> in adults, and low CD4+ count ≤750 cells/mm<sup>3</sup> and high CD4+ count >750 cells/mm<sup>3</sup> in children, in HIV-INFECTED only

<sup>#</sup> Pneumococcal conjugate vaccine (PCV13) vaccination status in younger children from available records

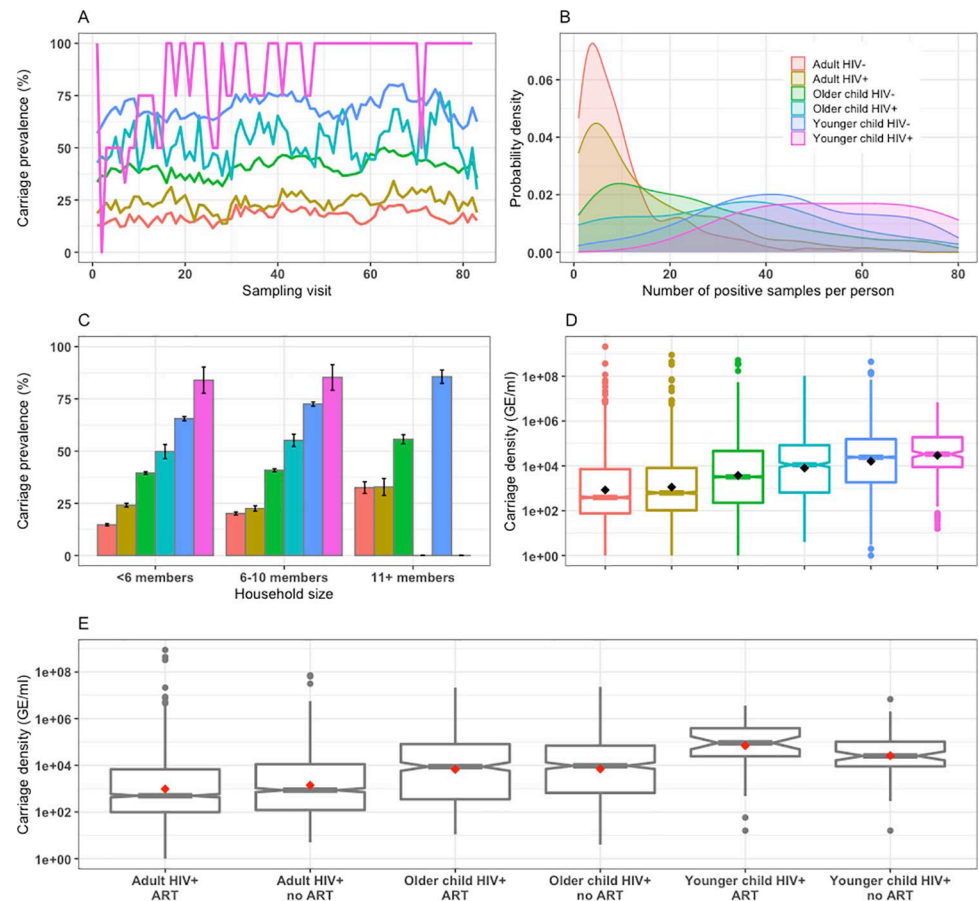
Standard deviation (SD)

N/A not applicable

<https://doi.org/10.1371/journal.pcbi.1009680.t001>

10.8–25.3% and HIV-uninfected female adults 10.2–24.0%, and between HIV-infected male adults 6.3–40.7% and HIV-infected female adults 12.3–34.6% (S5A Fig in [S1 File](#)).

Median pneumococcal carriage density, in genome equivalents per millilitre (GE/ml), was significantly higher in younger children (24,341, 95%CI: 22,638–26,122) than older children (3,490, 95%CI: 3,168–3,754) or adults (476, 95%CI: 429–522). Also, median carriage density



**Fig 2. Human immunodeficiency virus (HIV)-stratified pneumococcal carriage dynamics in younger children (<5 years-old), older children (5–17 years-old) and adults ( $\geq 18$  years-old) in South African households between 2016–2018.** Age and HIV-stratified pneumococcal carriage prevalence by different nasopharyngeal sampling visits (A), the likelihood of detecting pneumococcal carriage during visits (B), pneumococcal carriage prevalence by household sizes with 95% confidence intervals (CI) (C) and carriage densities with mean (black diamond), median and associated 95% CI of median, 25<sup>th</sup> and 75<sup>th</sup> percentiles, minimum and maximum, and outlier where carriage density is measured as genome equivalents per millilitre (GE/ml) (D). Age and antiretroviral therapy (ART) stratified carriage density with mean (red diamonds), median and associated 95%CI of median, 25<sup>th</sup> and 75<sup>th</sup> percentiles, minimum and maximum, and outlier (notched boxplot) (E).

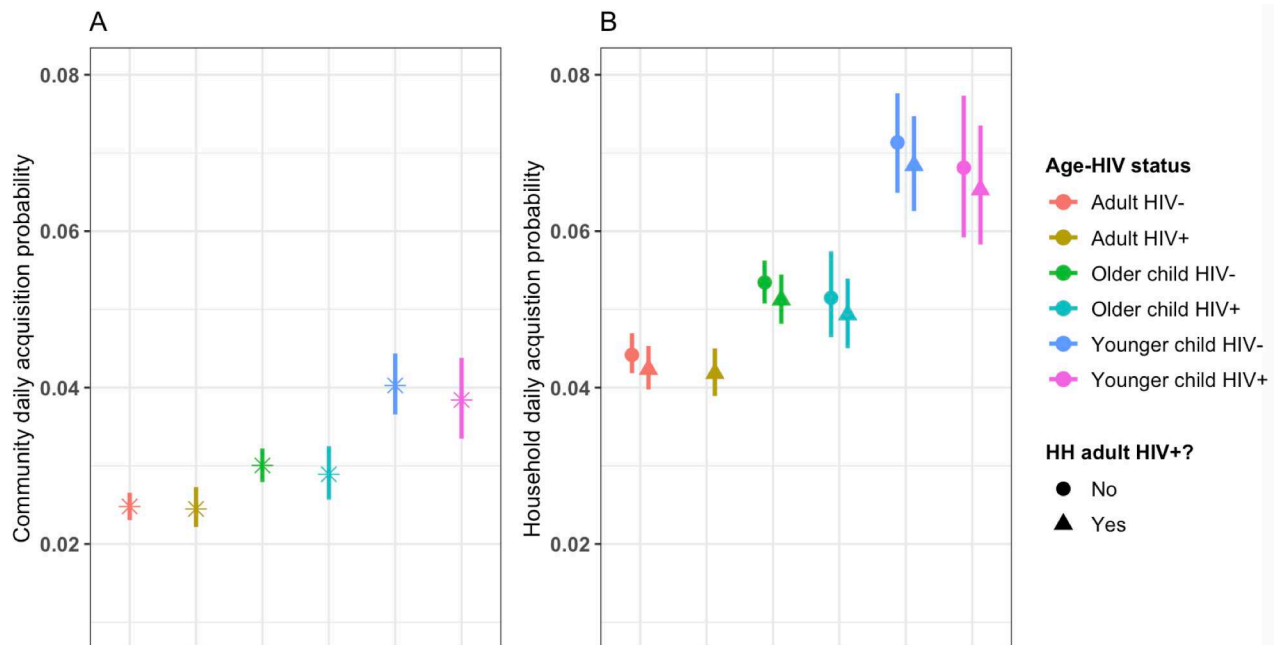
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was higher in HIV-infected than HIV-uninfected older children (11,156, 95%CI: 8,681–13,948 vs 3,221, 95%CI: 2,911–3,472) or adults (622, 95%CI: 507–714 vs 389, 95%CI: 311–435), and not in younger children (33,050, 95%CI: 22,690–42,293 vs 24,124, 95%CI: 22,547–25,838) (Fig 2D). Conversely, median carriage density was similar between those not on ART compared to those on ART in older children (9,624, 95%CI: 5,289–11,843 vs 8,818, 95%CI: 5,102–12,720) or adults (861, 95%CI: 508–1,001 vs 499, 95%CI: 382–586), and not in younger children (25,430, 95%CI: 13,138–40,245 vs 91,566, 95%CI: 43,355–265,628) (Fig 2E). About 14.4%, 95% CI: 13.7–15.0 of negative NP swab results were estimated probabilistically to be false negatives.

### Pneumococcal carriage acquisition

Overall, pneumococcal carriage acquisition was higher in older children (1.15, 95%CI: 1.08–1.23) and younger children (1.52, 95%CI: 1.38–1.68) than adults. Acquisition of carriage was more frequently observed when at least another household member was infected half a week





**Fig 3. Community and within household (HH) acquisitions of pneumococcal carriage in younger children (<5 years-old), older children (5–17 years-old) and adults ( $\geq 18$  years-old) in South African households between 2016–2018.** Age and human immunodeficiency virus (HIV) stratified estimates of community carriage acquisition probability per day (A) and within household carriage acquisition probability per day over the total follow-up period (B), comparing household without HIV-infected adult(s) (HIV-) to households with HIV-infected adult(s) (+).

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before (and hence attributed to household transmission) than in previously uninfected households (1.80, 95%CI: 1.68–1.93). Irrespective of age and HIV status, acquisition rates from within the household increased with household size; by 1.05 (95%CI: 1.00–1.10) in households with 6–10 members and by 1.41 (95%CI: 1.24–1.60) in households with 11 or more members compared to households with less than 6 members. However, within household carriage acquisition rates in children, irrespective of age group and HIV status, were not higher in the households with at least one HIV-infected adult (0.95, 95%CI: 0.91–1.01) (Fig 3 and Table 2). In addition, daily carriage acquisition rates in HIV-uninfected younger children did not significantly vary between households with HIV-infected female adults (0.14, 95%CI: 0.12–0.17) and those with HIV-infected male adults (0.13, 95%CI: 0.11–0.15) (S5 Fig in S1 File).

We estimated 3.8 carriage acquisition episodes per year, (95%CI: 3.4–4.2) for HIV-infected younger children, 5.9 (95%CI: 5.4–6.3) for HIV-uninfected younger children, 7.4 (95%CI: 6.7–8.1) for HIV-infected older children and 10.6 (95%CI: 10.2–11.0) for HIV-uninfected older children from households with at least one HIV-infected adult, and these were similar to their counterparts from households without HIV-infected adults (3.8, 95%CI: 3.3–4.2 and 5.8, 95%CI: 5.4–6.3, and 7.3, 95%CI: 6.6–8.0 and 10.3, 95%CI: 9.9–10.8, respectively) (Fig 3 and Table B in S1 File).

### Pneumococcal carriage duration

The average duration of pneumococcal carriage was highest in HIV-infected and HIV-uninfected younger children (107.9 days, 95%CI: 92.1–124.7 and 56.3 days, 95%CI: 51.1–62.1) followed by HIV-infected and HIV-uninfected older children (33.9 days, 95%CI: 29.9–38.6 and 17.9 days, 95%CI: 16.8–18.5), and HIV-infected and HIV-uninfected adults (11.4 days, 95%CI: 10.2–12.8 and 6.0 days, 95%CI: 5.6–6.3) (Fig 4C and 4D and Table C in S1 File).

Pneumococcal carriage cleared slower in older children (Hazard Ratio [HR]: 0.34, 95%CI: 0.31–0.36) and younger children (HR: 0.10, 95%CI: 0.09–0.12) when compared to adults. Carriage clearance was slower in HIV-infected than in HIV-uninfected individuals (HR: 0.52, 95%CI: 0.46–0.59), and faster in HIV-infected individuals with successfully suppressed viral load than in those without successful viral suppression (HR: 1.29, 95%CI: 1.13–1.47) (Fig 4B and Table 2). Antibiotic use may have accelerated pneumococcal clearance; however, the effect was not statistically significant (HR: 1.47, 95%CI: 0.67–3.25) (Fig 4A and Table 2).

### Sensitivity analysis

In the sensitivity analysis, a model that included age, HIV status, antibiotic use, and ART status as potential modifiers for pneumococcal carriage duration had the lowest AIC score as well as

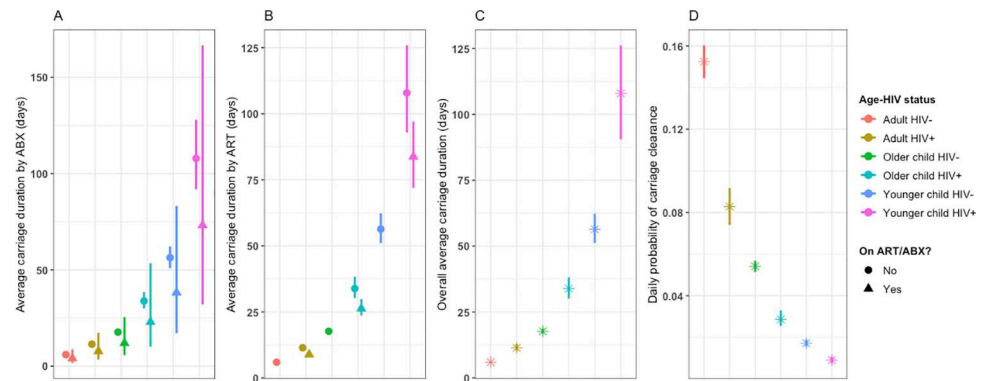
**Table 2. Effects of included covariates on pneumococcal acquisition and clearance rates estimated in the hidden Markov model.**

Description	Hazard Ratio (95%CI) <sup>‡</sup>
<u>Pneumococcal carriage acquisition</u>	
Age in years (y)	
Adult, ≥18y	Reference
Older child, 5–17y	1.15 (1.08–1.23)
Younger child, <5y	1.52 (1.38–1.68)
HIV status	
Negative	Reference
Positive	0.95 (0.87–1.04)
Children living with ≥1 HIV-infected adults	
No	Reference
Yes	0.95 (0.91–1.01)
Place of carriage exposure	
Community	Reference
Within household	1.80 (1.68–1.93)
Household size	
<6 members	Reference
6–10 members	1.05 (1.00–1.10)
11+ members	1.41 (1.24–1.60)
<u>Pneumococcal carriage clearance</u>	
Age in years (y)	
Adult, ≥18y	Reference
Older child, 5–17	0.34 (0.31–0.36)
Younger child, <5y	0.10 (0.09–0.12)
HIV status	
Negative	Reference
Positive	0.52 (0.46–0.59)
Antibiotic use	
No	Reference
Yes	1.47 (0.67–3.25)
Viral load-based ART status <sup>†</sup>	
Not on ART	Reference
On ART	1.29 (1.13–1.47)

<sup>†</sup> ART use status based on viral load results

<sup>‡</sup> 95% confidence interval (95%CI)

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**Fig 4. Duration of pneumococcal carriage in younger children (<5 years-old), older children (5–17 years-old) and adults (≥18 years-old) in South African households between 2016–2018.** Age and human immunodeficiency virus (HIV) stratified average carriage duration in days comparing individuals on antibiotics (ABX) (triangular shape) to those not on ABX (circular shape) (A), and individuals on antiretroviral therapy (ART) (triangle shape) to those not on ART (circle shape) (B). Age and HIV stratified overall average carriage duration in days (C). Age and HIV stratified daily probability of carriage clearance (D).

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for including both antibiotic use and ART status (Table A in [S1 File](#)). Increasing the number of HIV-infected adults within household to 1, 2, 3, 4, and 5 resulted in similar estimates of pneumococcal carriage acquisition in younger or older children (S3A Fig in [S1 File](#)). Our results were also robust when instead of assuming a time homogeneous hidden Markov model, we allowed for the estimation of time varying transition probabilities (S3B Fig in [S1 File](#))

## Discussion

We used a HMM to better understand pneumococcal carriage dynamics, and the role of HIV-infected adults in it, using data from a densely sampled longitudinal South African cohort using data from 115,595 nasopharyngeal swabs. We estimated that children have higher acquisition rates and duration of carriage than adults, and that, within a household, HIV-infected adults are not more likely to transmit pneumococci to children than HIV-uninfected adults. Pneumococcal acquisition events increased with larger household size irrespective of age and HIV status. Although ART use reduced pneumococcal carriage duration in HIV-infected children and adults, they still carry pneumococci for longer than their HIV-uninfected counterparts.

Heterogeneous household acquisition rates higher in children than adults have been reported previously [[34,51–55](#)], reflecting setting-specific population mixing behaviour and immunisation levels. Similarly, and for the first time in the presence of a mature infant PCV routine vaccination programme, we find that children both have higher acquisition rates than adults and carry pneumococci for longer, making them a likely key source for pneumococcal transmission in and beyond the household [[56,57](#)]. Moreover, adults simply have far lower carriage duration than children. So, even though HIV-infected adults have slightly longer carriage duration, the risk of carriage acquisition in children from an adult is far lower than from another child or in the community.

We postulated that HIV-infected adults were more likely to carry pneumococci and may have higher carriage density which individually or in combination may increase their risk for pneumococcal transmission compared to HIV-uninfected adults. Thus, carriage density was not controlled in this model for being on the causal pathway. Moreover, if carriage density indeed influences carriage transmission, then false negatives in low density carriers are

partially captured because individuals who are more transmissible are more likely to be correctly detected. Prior to infant PCV introduction, a study in Malawi showed that HIV-infected adults on ART had higher carriage prevalence than those not on ART [5], and two studies in South Africa also found that HIV-infected adults (mothers) had higher carriage prevalence than their HIV-uninfected counterparts, irrespective of ART status [34,51]. In addition, HIV-infected adults (mothers) were found to transmit pneumococci to their children more often than HIV-uninfected peers [34]. We generated additional evidence showing that, in the PCV era, carriage prevalence is slightly increased in HIV-infected adults on ART compared to HIV-uninfected adults as a result of reduced carriage clearance rates. We also show that median carriage density is higher in HIV-infected than HIV-uninfected adults. However, we find no evidence that carriage density is modified by ART status in HIV-infected adults (Fig 2). Further research may need to investigate whether differential effects of ART on pneumococcal carriage density in adults by country may be driven by types of ART regimens used.

Furthermore, our model estimates that the presence of an HIV-infected adult in the household does not increase the risk for pneumococcal carriage acquisition in co-habiting children. Although it is possible that there may have been other HIV-infected adults within households who were not enrolled into the study, it is unlikely this would alter the results given the insensitive acquisition estimates with increasing number of HIV-infected adults within household (S3A Fig in S1 File). These findings support the notion that ART largely, but not completely, reconstitutes the anti-pneumococcal mucosal immune response in HIV-infected adults [8]. This would imply that HIV-infected adults do not contribute disproportionately to pneumococcal transmission when on ART and hence that their vaccination is unlikely to substantially add to the herd protection already induced by the childhood immunisation programme although vaccination will provide direct protection against IPD in HIV-infected adults.

Our observation of increasing pneumococcal carriage acquisition rates with higher household size has also been reported previously [38] and suggests density dependent transmission in the household [58]. In line with evidence before infant PCV introduction [34,38], we find that pneumococcal carriage acquisition probabilities from the community were higher in children than in adults irrespective of HIV status, likely in part due to frequent effective contacts among playschool children [51,59] and immature immunity in children relatively to adults. We also estimate that children were twice as likely to get infected from within the household than from the community. However, we base this inference on the identified pneumococcal carriage in a household member at the previous visit. On the other hand, unlike previous household transmission models [34,38], our main model did not explicitly account for the number of other house members with carriage when estimating individual probability to carriage acquisition. However, it included a household size covariate to adjust for the contribution to carriage acquisition from housemates. Since pneumococcal infection rates usually increase with household size [38], this ensured that an individual living in a household with more members and likely to spread pneumococci has higher probability to carriage acquisition than smaller households with potentially fewer carrying individuals. A potential caveat could be if only a relatively small number of persons in a large household are indeed carrying which could overestimate infection contribution (S6 Fig in S1 File).

In the absence of serotyping of the pneumococcal isolates, our inferences may be prone to overestimation within household transmission by linking family members who in fact were infected with different pneumococcal serotypes. Similarly, serotyping would enhance our ability to differentiate a single and long carriage episode from almost immediate re-acquisition or the clearance of the dominant serotype while the previously subdominant serotype persists. This may have led to an overestimation of carriage duration and underestimated clearance rates. However, the mean carriage duration of 56 days (51–62) in HIV-uninfected children

estimated in this study aligns with studies that used serotype data [34,37,51,60]. While both the estimates for duration of carriage and the contribution of household transmission may be somewhat exaggerated, the lack of serotyping should not have affected our primary outcome, the relative contribution of HIV infected adults to pneumococcal transmission.

The use of ART, as inferred from measured viral load in study participants, reduced pneumococcal carriage duration by 22% compared to no ART use within each age group of HIV-infected participants. However, mean pneumococcal carriage duration remained slightly higher than their HIV-uninfected counterparts (Fig 4). Our model also estimated the sensitivity of the swabbing and qPCR testing regime for the detection of pneumococcal carriage. We estimate that about 1 in 7 swabs were misclassified as pneumococcal negative. False negatives might have arisen as a result of the sampling technique or if samples contained insufficient quantities of bacteria to successfully amplify and detect [58]. We assumed 100% specificity of an assay targeting the autolysin gene as the probability of false positives is seemingly very low [32,60]. However, lower specificity would yield slightly lower acquisition rates than estimated in this study. Our estimated misclassification probability in this study is within 10–20% range of values that were reported elsewhere [60,61].

In conclusion, we used one of the most densely sampled longitudinal pneumococcal carriage studies to infer the role of HIV-infected adults in pneumococcal transmission in the PCV and ART era. We find that the transmission risk from HIV-infected adults largely aligns with that of their uninfected counterpart. This implies that PCV use in HIV-infected adults who have access to ART would reduce their risk for pneumococcal disease but may have little added benefit over vaccinating other adults to the indirect protection against carriage of the rest of the population.

## Supporting information

**S1 File. Additional information on hidden Markov methods and model outputs.** Table A. Multistate model comparisons between hidden Markov models of specified degree of freedom ( $df$ ) using Akaike Information Criterion (AIC) after fitting the models to longitudinal pneumococcal carriage data in South African households, 2016–2018. Table B. Maximum likelihood parameter estimates and 95% confidence intervals (95%CI) for acquisition probabilities within household and from community from a hidden Markov model that is fitted to pneumococcal carriage data in South African households, 2016–18. Table C. Maximum likelihood parameter estimates and 95% confidence intervals (95%CI) for carriage duration and by antibiotic use and antiretroviral therapy (ART) from a hidden Markov model that is fitted to pneumococcal carriage data in South African households, 2016–18. S1 Fig. Hidden Markov model (HMM) convergence and predictions. HMM convergence estimated using maximum likelihood, given 5 Markov chains each with 1000 iterations. Each chain is a unique pair of initial infected ( $q_{12}$ ) and susceptible ( $q_{21}$ ) intensities converging to similar final baseline transition intensities, and  $2 \times \log$ -likelihood (A). HMM fitting assessment comparing the observed (diamond) to predicted (line) pneumococcal carriage and clearance with 95% predictive intervals of the model-fitted line, where observed data are grouped into 14-days intervals to compute fitted values (B). S2 Fig. The probabilities of the underlying states and the most likely path through them. Observed pneumococcal carriage results from NP swabs of two randomly selected persons A and B (first row). The underlying sequences of a fitted HMM given the observed sequence, found by the Viterbi algorithm through a recursive construction of the path with the highest probability (second row). For each observed infected state (first row), if the probability of the hidden infected state (third row) at each observation point is  $<100\%$  then it reflects misclassification. The probability of the hidden infected state is conditionally

on all the data and computed using Baum-Welch forward/backward algorithm. S3 Fig. Sensitivity analysis of varying covariate values in younger children (<5 years-old), older children (5–17 years-old) and adults ( $\geq 18$  years-old). The main analysis computed within household HIV+ and age-stratified acquisition per day comparing households with HIV + adult(s) to those without, whereas here, we compare households with 0,1,2,3,4 or 5 HIV + adult(s) (A). Similarly, the main analysis estimated acquisition probabilities for entire study follow-up period (0–289 days), whereas here, we estimate acquisition probability comparing samples collected between different periods. S4 Fig. HIV and Age distribution of study participants, and household size carriage acquisition dynamics in younger children (<5 years-old), older children (5–17 years-old) and adults ( $\geq 18$  years-old). Overall HIV and age distribution of study participants (A). HIV and age distribution of study participants by their household size (B). HIV and age-stratified carriage acquisition probability per day by household size (C). S5 Fig. Pneumococcal carriage prevalence in adults ( $\geq 18$  years-old) (A), household pneumococcal acquisition in younger children (<5 years-old) and older children (5–17 years-old) by HIV status of household female adults living with these children (B), and household pneumococcal acquisition in younger and older children by HIV status of household male adults living with these children (C). S6 Fig. The contribution to individual pneumococcal carriage acquisition by age groups and HIV status. Within household individual carriage acquisition probability estimated by accounting for infections through household size or daily contribution from household members. (DOCX)

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### Declaration

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### Disclaimer

The findings and conclusions in this publication are those of the authors and do not necessarily represent the official position of the United States Centers for Disease Control and Prevention, the NIHR or the Department of Health and Social Care.

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