The Rise and Fall of HIV in High-Prevalence Countries: A Challenge for Mathematical Modeling

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Abstract: Several countries with generalized, high-prevalence HIV epidemics, mostly in sub-Saharan Africa, have experienced rapid declines in transmission. These HIV epidemics, often with rapid onsets, have generally been attributed to a combination of factors related to high-risk sexual behavior. The subsequent declines in these countries began prior to widespread therapy or implementation of any other major biomedical prevention. This change has been construed as evidence of behavior change, often on the basis of mathematical models, but direct evidence for behavior changes that would explain these declines is limited. Here, we look at the structure of current models and argue that the common “fixed risk per sexual contact” assumption favors the conclusion of substantial behavior changes. We argue that this assumption ignores reported non-linearities between exposure and risk. Taking this into account, we propose that some of the decline in HIV transmission may be part of the natural dynamics of the epidemic, and that several factors that have traditionally been ignored by modelers for lack of precise quantitative estimates may well hold the key to understanding epidemiologic trends.

Introduction

Since the discovery of the first HIV/AIDS case more than three decades ago, enormous progress has been made in understanding, treating, and preventing the infection. For example, there is now solid evidence that antiretroviral therapy (ART) taken by either HIV-infected or uninfected individuals can prevent HIV transmission [1–4]. Nevertheless, the epidemic keeps surprising us. In particular, it has been noted that in many countries, notably in highly affected sub-Saharan Africa, but also in India, HIV prevalence is dropping fast [5]. In Zimbabwe, once one of the worst affected countries, HIV prevalence levels have halved over the past 12 years, dropping from 30% in 1998 to 15% in 2011 [6]. However, Zimbabwe is certainly not the first country to experience such an epidemiological change, as in east and central Africa, declines have been observed as early as the mid-1990s (Figure 1A) [7–9].

While these drops in HIV prevalence are dramatic, underlying changes in HIV incidence must have occurred several years earlier and more rapidly (Figure 1B). This is because with an average survival time of approximately ten years (in the absence of treatment), even a sudden complete interruption in transmission would translate into a gradual decline in prevalence. In Nairobi, changes in infection risk seemed to occur in high-risk populations first, while HIV prevalence in the general population was still stable or increasing. In 1986, over 80% of sex workers were HIV-infected, but this declined to less than 50% after 1997. Also, incidence dropped dramatically in these sex workers [10]. In these women, the per-protected-act rate of HIV infection fell over 4-fold between 1985 and 2005, and in a recent analysis of 2008–2011, HIV incidence was 2.2% in over 1,500 person-years of follow-up (McKinnon et al., submitted).

This rapid decline has occurred at a time when we still do not fully understand why HIV struck sub-Saharan Africa so exceptionally hard in the first place. At its peak, HIV prevalence in several countries in Africa exceeded more than 20% of the adult population, although there was a lot of heterogeneity both within and between countries [11]. The epidemic onset in sub-Saharan Africa was also often extremely rapid, with estimated prevalence doubling rates of as little as one year, implying that each HIV-infected person annually infects, directly or indirectly, another (susceptible) individual [12]. With transmissions apparently occurring over the entire infected period [13], this would seem to suggest high levels of the basic reproduction number $R_0$ (i.e., the average number of secondary cases generated by an infected individual, over the duration of his infection, when all contacts are uninfected). This, however, seems to conflict with what studies have shown about HIV transmission (in)efficiency. For example, some 50% of HIV-affected couples are still discordant when discovered, and during follow-up the annual seroconversion rate of the HIV-negative partner has ranged from 2% to 12% [14–16].

With an average—untreated—survival after HIV infection being approximately ten years, this would seem to suggest that, in the heterosexual population, $R_0$ cannot be much larger than one. With such a low $R_0$, generalized heterosexual epidemics either cannot occur or only spread slowly [17]. This is, in fact, also what happened outside of sub-Saharan Africa.

This discrepancy between high-prevalence sub-Saharan Africa countries and the rest of the world has sparked many theories and debates about what made Africa “special.” Indeed, there is still no full consensus about the reasons why various African countries...
Figure 1. Adult (15–49) HIV prevalence (A) and annual incidence (B) in Zimbabwe, Kenya, Malawi, and Mozambique, 1990–2009 [8].
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experienced severe generalized HIV epidemics while most of the rest of the world did not. Most of the hypotheses focus on high-risk sexual behavior such as rapid partner change and low condom use, aggravated by transmission co-factors, notably other sexually transmitted infections that are spread by the same high-risk sexual behavior. Undoubtedly, highly connected sexual networks generated by unprotected, often transactional, sex with many different, frequently concurrent, partners are common in Africa [18–27]. This is also shown by the high levels of conventional, bacterial sexually transmitted infections (STIs) as well as HSV-2 in sub-Saharan Africa, infections that have been implicated as transmission co-factors, although their exact role is still moot [28–34]. These risk factors have been incorporated in many mathematical models of the HIV epidemic, models that have also been used to interpret the recent decline in HIV transmission. Several modeling studies interpreted the fall in HIV transmission, therefore, as a decline in the very factors that made sub-Saharan Africa special.

However, as throughout history epidemics have always come and gone, even without interventions, we should consider the possibility that the decline in HIV transmission may also have “natural” causes. Here we discuss findings, both epidemiological and from the basic sciences, which are currently ignored in HIV transmission models and suggest that some of the dramatic decline in HIV incidence may be part of the natural dynamics of the HIV epidemic and might have occurred even in the absence of any major behavior change. To fully explore the impact of these findings, they should be incorporated in future modeling studies.

Current HIV Models, Common Assumptions, and Their Limitations

Mathematical modeling of infectious diseases is a highly developed science or technology, and software to accelerate the development and implementation of models is increasingly available [35–39]. For modeling complex situations that are difficult to capture in finite sets of (differential) equations, individual- or “agent-based” micro-simulation, such as used in STDSIM and SimulAIDS, is now well developed [40–42]. Advanced methods for fitting models to data and to incorporate uncertainty into inference and predictions have been used [43,44]. From the very beginning of the HIV epidemic, mathematical modelers, together with epidemiologists and statisticians, have been heavily involved in shaping our understanding of the infection and its transmission [45]. They have also been instrumental in the interpretation of observed trends, in planning of prevention strategies, in guiding data collection, and in developing scenarios for the future course of the epidemic [46–48].

Most models, irrespective of their implementation and mathematical detail, incorporate a constant per-sexual-contact transmission risk, which may depend on the phase of infection of infected partners, and STI cofactor effects that increase this risk. Given these parameter values, infection can spread through dynamic networks modeled with variable complexity. This constant risk assumption implies that “pre-emptive saturation” is the only density-dependent mechanism at play [49]. That is, transmission will continue to increase until so many sex partners of HIV-infected individuals are already infected that (on average) each HIV-positive individual only infects one still uninfected individual. The latter condition defines endemic equilibrium, which will set in and remain “forever” unless behavior change or interventions shift this equilibrium to lower levels. Moreover, in homogeneous populations this equilibrium will be reached after a continuously increasing prevalence with perhaps a minor “overshoot,” as mortality lags—by about a decade—behind incidence [50]. Thus, when incidence declines, the common fixed risk per sexual contact assumption implies the conclusion of substantial sexual behavior changes.

As in various countries the decline in HIV transmission started prior to the year 2000, major biomedical interventions such as antiretroviral therapy (ART) or male circumcision campaigns cannot explain the decline in HIV transmission. Most current models thus necessarily interpret the epidemiological change as evidence of (large) changes in sexual behavior or depletion (by higher AIDS mortality rates) of the group with the highest rates of partner change or as proof of success of behavioral interventions [51–54]. Several modeling studies have indeed come to this conclusion, although it is unclear whether direct empirical studies, mostly self-reported and therefore potentially biased, support commensurate declines in risk behavior [55–69]. While such behavior changes undoubtedly contributed to the decline in transmission [70], they may not be the only explanation. That other factors may play a role is supported by the observation that population behavior change appears to fail to explain the early, rapid declines in risk per unprotected contact among Kenyan sex workers [10].

Non-linearity of the Relationship between the Number of HIV Exposures and Infection Risk

It is widely believed that HIV transmission is very inefficient. On average, a very small proportion of sexual encounters between infected and uninfected partners lead to transmission [71]. When transmission does take place, infection is often the result of a clonal expansion of a single founder virus, and the probability of any one virion (virus particle) successfully infecting a person must be on the order of one in a billion [72]. This is corroborated by SIV (monkey HIV) infection experiments, which often show a single cluster of infected cells in the female genital tract in the first few days of SIV infection [73]. Special factors, such as an STI cofactor effect, therefore have to be invoked to plausibly explain and model the observed epidemics.

Potentially misleading, however, may be the concept of per-contact transmission risk. Several observations disprove a constant per-coitus risk of transmission [74,75]. Downs and Vincenzi, among the first authors to study this, have shown the relationship between the number of sexual contacts and risk of HIV transmission to be highly non-linear. In their study involving 563 heterosexual partners of HIV-infected subjects, the risk of transmission was 10% for those with less than ten unprotected contacts and increased to only 23% after 2,000 unprotected contacts [76,77]. Thus, the first few sexual exposures are clearly more dangerous than subsequent ones. Sex worker contacts of Thai military conscripts were also associated with high per-contact risk, as were contacts among adult film actors [78–80]. Low-dose mucosal SIV challenge experiments in non-human primates suggest a similar phenomenon; while some monkeys become infected quite rapidly, others require far more challenges, at times with higher virus doses, to achieve infection [81]. It is unclear, however, how to unambiguously interpret this non-linearity with major implications for modeling and model predictions. Interestingly, another transmission non-linearity exists in the relationship between viral load and the risk of infection at each sexual contact. Each \(10^{log} \) increase in plasma HIV-1 RNA increases the per-act risk of transmission by 2.9-fold; that is, the transmission risk seems to be approximately proportional to the square root of the viral load [82]. A similar non-linear relationship appears to exist between HIV-1 RNA in the genital tract and the risk of transmission to an uninfected partner [83].
Perhaps, if (cumulative) risk is determined by numbers of virions to which a person is exposed, irrespective of the number of coital acts, then the non-linearity between number of contacts and transmission risk may well reflect the same biological phenomenon as between per-coitus viral dose and transmission risk.

**Biological Mechanisms of the Non-linearity in HIV Transmission**

Below we present some mechanisms that may account for the non-linearity in the relationship between exposure and transmission risk.

**Susceptibility of the HIV-negative partner**

As with everything biological, susceptibility to infection is likely to vary both among and within individuals. Early in the HIV epidemic, it was discovered that individuals homozygous for the Δ32 mutation of the CCR5 gene were practically resistant to HIV infection [84]. Other genetic factors also appear to play a role in susceptibility and resistance, but much of the genetics and biology of resistance have remained elusive, and behavior and acquired factors may also be important [85–89]. This raises the issue of whether the increased transmission during acute infection is totally due to an increased infectivity of the infected partner. Acute infection in the HIV-positive partner is often the first exposure of the negative partner to HIV, and it seems logical that the most susceptible individuals get infected first and fast.

In statistical terms, when such heterogeneity is taken into account, the duration to infection (in terms of exposure) is a mixture of exponential distributions, with characteristically declining hazard rates; i.e., risk appears to decline with the total number of previous exposures [90]. As people can seroconvert only once, heterogeneity in susceptibility is notoriously hard to measure directly, but studies are highly suggestive that it can be substantial. Sex workers in Nairobi in the 1980s reached very high (>80%) infection levels, but the risk of seroconversion was strongly negatively associated with duration of exposure and there was even evidence of a small, highly resistant subgroup [91]. In our most recent analysis, this decrease in risk was 24% per year of exposure (McKinnon et al., submitted). The same has been observed in other sex worker cohorts [92]. Whether heterogeneity in susceptibility between individuals offers a totally satisfactory explanation of the rapid HIV epidemiological change is still unclear. By itself, it seems to fail to satisfactorily explain the onset of the epidemic. It may be of interest that for the same final size of the epidemic, heterogeneity in susceptibility could give rise to faster epidemic onset. Perhaps because of the difficulty of obtaining empirical estimates of the level of heterogeneity in susceptibility, this variable has usually been ignored in HIV epidemic models. Yet, heterogeneity in susceptibility and resistance can have a major effect on the course of the epidemic. In fact, models that incorporate this heterogeneity predict that prevalence peaks and then drops, as actually observed (Figure 2) [50]. Heterogeneity in susceptibility in its most extreme form, where only a subset of individuals are (highly) susceptible, and the rest minimally susceptible, may also lead to a fragmentation of sexual networks as all connections (edges) via non-susceptibles (nodes) are effectively removed. Thus, sexual networks consisting of loosely connected “cliques” would fragment into disjointed subnetworks, with obvious implications for HIV spread. Heterogeneity in susceptibility over time, within individuals, almost certainly also exists but, except in the context of STIs, has remained largely elusive.

**Immunity**

Current mathematical models of HIV ignore the possibility of naturally acquired immunity. There is evidence of (cellular and other) immune responses to HIV among exposed but uninfected individuals, some of which are associated with reduced infection risk, but whether such responses are evidence of (long-term) protective acquired and/or adaptive immunity is unclear [93–97]. The success of the RV144 vaccine trial in Thailand should also alert us to the possible existence of forms of protective immunity to HIV [98,99]. Immunity could develop, for example, after exposure to defective virions, i.e., virus particles unable to establish a reproductive infection. Given these data, the non-linearity of the relationship between number of contacts and transmission risk may also be due to such acquired immunity, including mucosal immune mechanisms such as HIV target cell availability, for example [100]. If so, this immunity perhaps may only offer protection against one of several strains, or subtypes (clades) of the virus, with implications for the risks of partner change [101].

Another form of immunity that may be relevant is alloimmunity, i.e., immunity to (HLA) antigens expressed by sex partners. Early studies demonstrated potent anti-HIV activity induced by allo-immunization in humans [102]. There is evidence that transmission often takes place by HIV hitching a ride in cells. As such cells express HLA molecules of the host, alloimmune responses may be protective [103]. Evidence, albeit somewhat conflicting, for the importance of these responses comes from both mother-to-child and sexual transmission studies [104–108].

With the high strain diversity of HIV and the extremely high degree of polymorphism of HLA antigens, one would expect regular partners of HIV-infected individuals to benefit much more from these types of immunity than (say) sex workers whose immune systems have to cope with widely divergent HIV strains and HLA antigens and whose alloimmune responses may even be compromised by their high levels of concurrency [109]. If the non-linearity of risk is due to some form of specific immunity, then this would imply much greater risks associated with multiple and/or concurrent partnerships than if it were due to, for example, innate, broadly protective, susceptibility factors. New HIV-positive partners would entail a very high risk during the first couple of contacts, whereas innate resistance or susceptibility would not discriminate between contacts with the same and with different HIV-infected partners.

Interestingly, with alloimmunity, the risk associated with acute phase infection of a seroconverting regular partner would be much lower than the risk associated with a new seroconverting partner, as it (the alloimmunity) is acquired before HIV exposure. This would not be the case with immunity against the virus itself. It seems likely that this may affect the relative importance of concurrent versus new partnerships. To our knowledge, no study has explored the implications of these types of immunity for the spread of HIV. Infections for which immunity is a density-dependent mechanism, such as measles, are often characterized by epidemiological patterns of (periodic) violent outbreaks, followed by subsequent declines when high levels of immunity in the population (herd immunity) form an impediment to ongoing transmission. Therefore, the epidemiological implications of HIV immunity merit further study.

**Heterogeneity in infectiousness**

As viral load varies among people, one can expect some individuals to be greater shedders of infectious virions, and thereby more infectious, than others. Early-stage, acute infection has been associated with high viral loads and high infectiousness, and this seems to be the preferred “consensus” interpretation of the
non-linearity of the relationship between exposure and risk. This acute infection period (first few months) has been estimated to be up to 26 times as infectious as the subsequent stage of infection [13,110]. Yet, paradoxically, the elevation in transmissibility observed during acute infection appears to be greater than would be expected based on viral load alone (although not all virions may be equally infectious), and direct evidence implicating acute infections only comes from a single study that ignored alternative explanations and whose methodology has been questioned [111–113]. Nevertheless, several mathematical models have incorporated this aspect of transmission and even made quite definitive estimates about the role of acute infections in overall HIV transmission [114]. Unfortunately, the methodology used for estimating this increased transmissibility completely ignores factors such as heterogeneity in susceptibility (see above). Proper estimation of the risk associated with acute infection per se would require a comparison between new, HIV-naïve partners of HIV-infected individuals in the acute and post-acute phase, which has never been done. Other, less understood, sources of heterogeneity in infectiousness also appear to exist. In a cross-sectional study from South Africa, viral loads varied by a factor of more than 10,000, but the factors underlying this heterogeneity are not completely clear [115]. Some factors have been recognized; for example, HLA-B57 and HLA-B27 genotypes have been associated with better viral control, and this leads one to expect that individuals with these HLA types are, on average, less infectious than others [116]. However, there is also within-person variation, and individuals may temporarily shed more infectious virions than usual [117]. Some anecdotal “superspreading” events suggest that heterogeneity in infectiousness can be substantial. A Zairian man in Belgium infected 11 out of 19 female sex partners, often after a single sexual contact [118]. Another case infected at least 11 of his many female sex partners [119]. It is unclear what this implies for HIV transmission in general, or the difference between transmission in Africa and other parts of the world. Africa has higher rates of tropical infections, including malaria, which may increase HIV viral load and transmission [120,121]. Heterogeneity in infectiousness, as with heterogeneity in susceptibility, would explain why some exposed individuals seroconvert after a few sexual contacts, while others never acquire infection.

Conclusions

We have identified several aspects of HIV transmission that may provide additional, alternative reasons for the rise and fall of HIV in high-prevalence countries in Africa. Current models may be misleading because factors such as heterogeneity in susceptibility, which may have substantially impacted the course of the HIV epidemic, have been ignored due to a lack of precise quantitative estimates. Key observations, such as non-linearity of the relationship between the number sex contacts and HIV transmission and the high risks associated with a seroconverting partner have consistently been interpreted, perhaps misinterpreted, in only a single way, while other explanations, with different epidemiological impacts, have been discounted. Thus, when incidence declines, the common assumption of fixed risk per sexual contact

Figure 2. The projected course of the HIV epidemic (prevalence) under simplified assumptions about heterogeneity in susceptibility to HIV infection [50]. For comparison, the dashed curve shows the projected course in the absence of such heterogeneity, but with similar final size of the epidemic. doi:10.1371/journal.pcbi.1003459.g002
favors the conclusion of substantial changes in average sexual behavior.

What would our ideas imply for modeling? First, uncertainty in model structure, more than uncertainty in parameter values, dominates model uncertainty, and modelers should be aware of qualitatively different interpretations of available data. Microsimulation can be used to effectively model complex, non-linear processes, in contrast to most models consisting of sets of differential equations. Modelers should explore alternative interpretations of available data and, together with epidemiologists and basic scientists, try to formulate empirical studies to discriminate between these interpretations. Scientific journals and reviewers, rather than encouraging conformity, should encourage and invite debate and dissenting interpretations of epidemiological studies. Understanding HIV’s rise and fall could offer important lessons as to this and future epidemics unravel.

References

1. Grant RM, Lama JR, Anderson PL, McMahan V, Liu AY, et al. (2010) Preexposure chemoprophylaxis for HIV prevention in men who have sex with men. N Engl J Med 363: 2587–2599.

2. Cohen MS, Chen YQ, McCauley M, Gamble T, Hosseinipour MC, et al. (2011) Prevention of HIV-1 infection with early antiretroviral therapy. N Engl J Med 365: 398–406.

3. Baeten JM, Donnell D, Nduku P, Maquie NR, Campbell JD, et al. (2012) Antiretroviral prophylaxis for HIV prevention in heterosexual men and women. N Engl J Med 367: 399–410.

4. Tanser F, Barnighausen T, Grapsa E, Zaidi J, Newell ML (2013) High coverage of ART associated with decline in risk of HIV acquisition in rural KwaZulu-Natal, South Africa. Science 339: 966–971.

5. Arora P, Kumar R, Bhattacharya M, Nagelkerke N, Jha P (2008) Trends in HIV incidence in India from 2000 to 2007. Lancet 372: 289–290.

6. Batenman C (2011) HIV prevalence in Zimbabwe dropping like a stone. S Afr Med J 101: 10–11.

7. Mulder D, Nunn A, Kamali A, Kengeya-Kayondo J (1995) Decreasing HIV-1 seroprevalence in young adults in a rural Ugandan cohort. Br Med J 313: 835–838.

8. UNAIDS (2010) Report on the Global AIDS Epidemic: 2010. Available: http://www.unaids.org/globalreport/global_report.htm. Accessed 3 January 2014.

9. Hogan DR, Zaslavsky AM, Hammitt JK, Salomon JA (2010) Flexible epidemiological model for estimates and short-term projections in generalised HIV/AIDS epidemics. Sex Transm Infect 86: i48–i52.

10. Kimani J, Kaul R, Nagelkerke NJD, Luo M, MacDonald KS, et al. (2008) Reduced rates of HIV acquisition during protected sex by Kenyan female sex workers precluding population declines in HIV prevalence. AIDS 22: 131–137.

11. Piot P, Bartos M, Grispe E, Zaidi J, Newell ML (2001) The global impact of HIV/AIDS. Nature 410: 968–973.

12. Williams B, Campbell C (1998) Understanding the epidemic of HIV in Sub-Saharan Africa. Analysis of the antenatal clinic survey data. S Afr Med J 88: 247–251.

13. Williams B, Campbell C (1998) Understanding the epidemic of HIV in Sub-Saharan Africa: A systematic review and meta-analysis. Lancet Infect Dis 10: 710–770.

14. Cheman knowing H, Cimoni L, Shetron J, Hallett TB, Abu-Raddad LJ (2012) Distinct HIV discordancy patterns by epidemic size in stable sexual partnerships in sub-Saharan Africa. Sex Transm Infect 88: 51–57.

15. The World Bank Group (2011) Emerging Issues in Today's HIV Research: Debate 3. Discordant Couples and HIV Transmission. Available: http://www.wbgroup.org/2011/Debate_3_Summary_R3_tagged.pdf. Accessed 3 January 2014.

16. The World Bank Group (2011) World Bank Debate Series: Debate 4. Concurrent Sexual Partnerships. Available: http://siteresources.worldbank.org/INT/HIV/AIDS/Resources/1357598-129872814065987/Debate_4SUMMARYConcurrentSexualPartnerships.pdf. Accessed 3 January 2014.

17. Rothenberg R (2009) HIV transmission networks. Curr Opin HIV AIDS 4: 260–263.

18. Vauthier J, Leyerl R, Dalabetta G, Crabbe F, Alary M, et al. (2006) Estimates of the number of female sex workers in different regions of the world. Sex Transm Infect 82: iii30–iii35.

19. Chen L, Jha P, Stirling R, et al. (2007) Sexual risk factors for HIV infection in early and advanced HIV epidemics in sub-Saharan Africa: systematic overview of 68 epidemiological studies. PLOS ONE 2: e1001. doi:10.1371/journal.pone.0001001.

20. Fleming DT, Wasserheit JN (1999) From epidemiological synergy to public health policy and practice: the contribution of other sexually transmitted diseases to sexual transmission of HIV infection. Sex Transm Infect 75: 3–17.

21. Gl små RM, Mayaud P, Arin A, Mabey D, Rosensiz C, et al. (2012) Prevalence of malaria and sexually transmitted and reproductive tract infections in pregnancy in sub-Saharan Africa: a systematic review. J Am Med Assoc 307: 2070–2076.

22. Smith JS, Robinson NJ (2003) Age-specific prevalence of infection with herpes simplex virus types 2 and 1: a global review. J Infect Dis 188: S3–S28.

23. Koerner R, Lee V, Njagura NJ, Lhambenje JDF (2001) Estimating the magnitude of STD cofactor effects on HIV transmission: how well can it be done? Sex Transm Infect 28: 615–621.

24. Kaul R, Kimani J, Nagelkerke NJD, Fonk K, Njupi EN, et al. (2004) Monthly antibiotic chemoprophylaxis and incidence of sexually transmitted infections and HIV-1 infection in Kenyan sex workers: a randomized controlled trial. J Am Med Assoc 291: 2535–2540.

25. Kamali A, Quigley M, Nakinyingi J, Kinman J, Kengeya-Kayondo J, et al. (2003) Syndromic management of sexually-transmitted infections and behaviour change interventions on transmission of HIV-1 in rural Uganda: a community randomised trial. Lancet 361: 645–652.

26. Wawer MJ, Sewankambo NK, Serwadda D, Quinn TC, Paxton LA, et al. (1999) Control of sexually transmitted diseases for AIDS prevention in Uganda: a randomised community trial. Rakai Project Study Group. Lancet 353: 352–355.

27. Anderson RM, May R (1991) Infectious diseases of humans: Dynamics and control. Oxford: Oxford University Press.

28. Kramer A, Kretzmar M, Kriekkev K (2009) Modern infectious disease epidemiology: Concepts, methods, mathematical models, and public health. New York: Springer.

29. Vynnycky E, White R (2010) An introduction to infectious disease modelling. Oxford: Oxford University Press.

30. Diekmann O, Heesterbeek JAP (2000) Mathematical Epidemiology of Infectious Diseases: Model Building, Analysis and Interpretation. New York: John Wiley & Sons Ltd.

31. Macey RJ, Oster GF (2001) Berkeley Madonna: Modelling and Analysis of Dynamic Systems. Available: http://www.berkeleymadonna.com/. Accessed 3 January 2014.

32. Koerner R, Van Vliet C, Grosskurth H, Gavyole A, Van der Ploeg CP, et al. (2000) Model-based evaluation of single-round mass treatment of sexually transmitted diseases for HIV control in a rural African population. AIDS 14: 573–592.

33. Robinson NJ, Mulder D, Auvert B, Whitworth J, Hayes R (1999) Type of partnership and heterosexual spread of HIV-1 infection in rural Uganda: results from simulation modelling. Int J STD AIDS 10: 718–725.

34.IMA (2005) International Microsimulation Association (IMA). Available: http://www.microsimulation.org. Accessed 1 January 2014.

35. Raftery AE, Bao L (2010) Estimating and projecting trends in HIV/AIDS generalized epidemics using incremental mixture importance sampling. Biometrics 66: 1162–1173.

36. Bartos M, Jha P, Stirling R, et al. (2007) A stochastic infection rate model for estimating and projecting national HIV prevalence rates. Sex Transm Infect 86: iv30–iv35.

37. Johnson LF, White PJ (2011) A review of mathematical models of HIV/AIDS. Oxford: Oxford University Press.

38. Robinson NJ, Bao L, Ghan P, Whitworth J, Hayes R (1999) Type of partnership and heterosexual spread of HIV infection in rural Uganda: results from simulation modelling. Int J STD AIDS 10: 718–725.

39. Raftery AE, Bao L (2010) Estimating and projecting trends in HIV/AIDS generalized epidemics using incremental mixture importance sampling. Biometrics 66: 1162–1173.

40. Raftery AE, Bao L (2010) A stochastic infection rate model for estimating and projecting national HIV prevalence rates. Sex Transm Infect 86: iv30–iv35.

41. Robinson NJ, Bao L, Ghan P, Whitworth J, Hayes R (1999) Type of partnership and heterosexual spread of HIV infection in rural Uganda: results from simulation modelling. Int J STD AIDS 10: 718–725.

42. IMA (2005) International Microsimulation Association (IMA). Available: http://www.microsimulation.org. Accessed 1 January 2014.

43. Raftery AE, Bao L (2010) Estimating and projecting trends in HIV/AIDS generalized epidemics using incremental mixture importance sampling. Biometrics 66: 1162–1173.

44. Raftery AE, Bao L (2010) A stochastic infection rate model for estimating and projecting national HIV prevalence rates. Sex Transm Infect 86: iv30–iv35.

45. Johnson LF, White PJ (2011) A review of mathematical models of HIV/AIDS. Oxford: Oxford University Press.

46. Robinson NJ, Bao L, Ghan P, Whitworth J, Hayes R (1999) Type of partnership and heterosexual spread of HIV infection in rural Uganda: results from simulation modelling. Int J STD AIDS 10: 718–725.
56. Gallo MF, Behets FM, Steiner MJ, Thomsen SC, Ombidi W, et al. (2007)
55. Gallo MF, Behets FM, Steiner MJ, Hobbs MM, Hoke TH, et al. (2006)
49. Ward H (2007) Prevention strategies for sexually transmitted infections:
74. Powers KA, Poole C, Pettifor AE, Cohen MS (2008) Rethinking the
73. Li Q, Estes JD, Schlievert PM, Duan L, Brosnahan AJ, et al. (2009) Glycerol
72. Keele BF, Giorgi EE, Salazar-Gonzalez JF, Decker JM, Pham KT, et al. (2008)
70. Stoneburner RL, Low-Beer D (2004) Population-level HIV declines and
68. Paul-Ebhohimhen VA, Poobalan A, van Teijlingen ER (2008) A systematic
67. Schmidt WP, Van Der Loeff MS, Aaby P, Whittle H, Bakker R, et al. (2008)
62. Deuchert E (2011) The virgin HIV puzzle: can misreporting account for the
61. Morris M (1993) Telling tails explain the discrepancy in sexual partner reports.
58. Mapingure MP, Msuya S, Kurewa NE, Munjoma MW, Sam N, et al. (2010)
57. Ferrer B, Caraël M, Buve A, Auvert B, Laourou M, et al. (2001) Comparison of
55. Rahmouni Z, Cmarko T, Halpern L, Castle L, Sow S, et al. (2005) Sexual
53. Stein P, Morin C, Nkengasong JN, Embali M, Chibale K, et al. (2004)
52. Sage P, Weller S, Corbane C, Huyghe B, De Vranken J, et al. (2004)
51. Poirel L, Nordmann P, Grothe C, et al. (2004) Analysis of the plasmid pTB132
50. Mzimeka SA, Coutsoudis A, et al. (2003) Prevalence and correlates of
49. Nkengasong JN, Sow S, Njim S, et al. (2005) Longitudinal assessment of
48. Rerks-Ngarm S, Pitisuttithum P, Nitayaphan S, et al. (2009) Vaccination with
47. Fowke KR, Haynes BF, McElrath MJ, Darrow CC, et al. (2002) HLA-DQ alleles
46. van de Perre P, Desloges V, et al. (1989) Antiretroviral treatment in patients
45. Bello-Macaulay V, Fofana A, Dacosta A, et al. (2005) Four-year trends in
44. MacKernan G, Ntoumi F, Blendis L, et al. (2000) HIV-1 and HIV-2 infection and
43. Dallabetta GA, Grant RM, et al. (1991) Hypervariable region IV (HIV-1, HVR-IV)
42. Abdul-Muhsin M, Thiermann H, et al. (2005) The role of sexual and drug
41. Mazzotta A, Pellegrino P, et al. (2003) The HLA-DQb1 genotypes and their
40. Hayes VE, van der Groen K, et al. (2003) Sequence and phylogenetic analysis
39. Ramachandran S, Arakiteshwar P, et al. (2005) HLA-B7 and HLA-B27
38. Wadsworth J, et al. (2003) Evaluation of repeat HIV serosurveillance
37. Kimani J, Kaul R, Ngugi E, et al. (2001) Determinants of per-coital-act HIV-1 infectivity among African HIV-1-serodiscordant couples. J Infect Dis 205: 338-346.
36. Rerks-Ngarm S, Pitisuttithum P, Nitayaphan S, et al. (2009) Vaccination with
35. Soudijn M, van de Poll F, et al. (2003) Specific CD4 and CD8 T cell responses to
34. Dallabetta GA, Grant RM, et al. (1991) Hypervariable region IV (HIV-1, HVR-IV)
33. Mazzotta A, Pellegrino P, et al. (2003) The HLA-DQb1 genotypes and their
32. Caskey CT, et al. (1987) A genetic linkage map of man.
31. Bollinger RC, et al. (1988) DNA sequence of the human major histocompatibility
30. Hirsch MS, et al. (1989) Human T cell line established from a patient infected
29. Kumar KS, et al. (2002) A family study of tuberculosis and HIV-1 infection in
28. Haynes BF, et al. (2008) Influence of HLA-DQ class II alleles and haplotypes and their associations with resistance and susceptibility to HIV-1 infection. AIDS 22: 807-816.
27. Rerks-Ngarm S, Pitisuttithum P, Nitayaphan S, et al. (2009) Vaccination with
26. Rerks-Ngarm S, Pitisuttithum P, Nitayaphan S, et al. (2009) Vaccination with
25. Tietze U, et al. (2003) Analysis of the Nef gene of HIV-1 in 827 patients from South
24. Biggar RJ, et al. (1995) A high proportion of HIV-1-infected individuals are seronegative for HIV-1 antibodies.
23. Haynes BF, et al. (2008) Influence of HLA-DQ class II alleles and haplotypes and their associations with resistance and susceptibility to HIV-1 infection. AIDS 22: 807-816.
22. arrange for a family study of tuberculosis and HIV-1 infection in patients from South Africa.
21. Fowke KR, Haynes BF, McElrath MJ, Darrow CC, et al. (2002) HLA-DQ alleles
20. Fowke KR, Haynes BF, McElrath MJ, Darrow CC, et al. (2002) HLA-DQ alleles
19. Bautista I, et al. (1996) Heterogeneity in host HIV susceptibility as a potential contributor to recent decreases in HIV prevalence in Africa. AIDS 20: 125-130.
18. Weller S, Moyal L, et al. (2005) Quantifying ongoing HIV-1 exposure in HIV-1-serodiscordant couples with quantified exposure. PLOS ONE 6: e29632. doi:10.1371/journal.pone.0029632
17. Mackelprang RD, Baeten JM, Donnell D, Celum C, Farquhar C, et al. (2012) Quantifying ongoing HIV-1 exposure in HIV-1-serodiscordant couples to identify individuals with potential host resistance to HIV-1. J Infect Dis 206: 1299–1308.
16. Joshi B, et al. (2003) Identification and characterization of transmitted and early founder virus sequences in plasma from HIV-1 serodiscordant couples. J Gen Virol 84: i43–i49.
15. Kaul R, Rutherford J, Rowland-Jones SL, Kimani J, Onyango J, et al. (2004) HIV-1 Env-specific cytotoxic T-lymphocyte responses in exposed, uninfected Kenyan sex workers: a prospective analysis. AIDS 18: 2067–2089.
14. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
13. Kaul R, Rowland-Jones SL, Kimani J, Fowke K, Dong T, et al. (2001) New insights into HIV-1-specific cytotoxic T-lymphocyte responses in exposed, persistently seronegative Kenyan sex workers. Immunol Lett 79: 125–131.
12. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
11. Kaul R, Rowland-Jones SL, Kimani J, Fowke K, Dong T, et al. (2001) New insights into HIV-1-specific cytotoxic T-lymphocyte responses in exposed, persistently seronegative Kenyan sex workers. Immunol Lett 79: 125–131.
10. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
9. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
8. Kaul R, Rowland-Jones SL, Kimani J, Fowke K, Dong T, et al. (2001) New insights into HIV-1-specific cytotoxic T-lymphocyte responses in exposed, persistently seronegative Kenyan sex workers. Immunol Lett 79: 125–131.
7. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
6. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
5. Fowke KR, Kaul R, Rosemthal KJ, Ouyang J, Kimani J, et al. (2000) HIV-1-specific cellular immune responses among HIV-1-infected sex workers. Immunol Cell Biol 78: 586–595.
102. Wang Y, Tao L, Mitchell E, Armstrong P, Vaughan R, et al. (1999) Alloimmunization elicits CD8+ T cell-derived chemokines, HIV suppressor factors and resistance to HIV infection in women. Nat Med 5: 1004–1009.
103. Anderson DJ, Politch JA, Nauditski AM, Blaskevitz CD, Podley J, et al. (2010) Targeting Trojan Horse leukocytes for HIV prevention. AIDS 24: 163–167.
104. MacDonald KS, Embree J, Nyenga S, Nagelkerke NJ, Gatia I, et al. (1998) Mother-child class I HLA concordance increases perinatal human immunodeficiency virus type 1 transmission. J Infect Dis 177: 551–556.
105. Laucher MA, Choy G, Embree JE, Nagelkerke NJ, Bwayo JJ, et al. (1998) Anti-HLA alloantibody is found in children but does not correlate with a lack of HIV type 1 transmission from infected mothers. AIDS Res Hum Retroviruses 14: 99–107.
106. Peters B, Whittall T, Babahamody K, Gray K, Vaughan R, Lehner T (2004) Effect of heterosexual intercourse on mucosal alloimmunisation and resistance to HIV-1 infection. Lancet 363: 518–524.
107. Leith JG, Clark DA, Matthews TJ, Rosenthal KL, Luscher MA, et al. (2003) Assessing human alloimmunization as a strategy for inducing HIV type 1 neutralizing anti-HLA responses. AIDS Res Hum Retroviruses 19: 957–965.
108. Dorak MT, Tang J, Perman-Aguilar A, Wentfall AO, Zulu I, et al. (2004) Transmission of HIV-1 and HLA-B allele-sharing within serodiscordant heterosexual Zambian couples. Lancet 363: 2137–2139.
109. Jennes W, Evertse D, Borget MY, Vuyyulete B, Maurice C, et al. (2006) Suppressed cellular alloimmune responses in HIV-exposed seronegative female sex workers. Clin Exp Immunol 143: 435–444.
110. Pichler CD, Tien HC, Eron JJ Jr, Vernazza PL, Leo SY, et al. (2004) Brief but efficient: acute HIV infection and the sexual transmission of HIV. J Infect Dis 189: 1785–1792.
111. Cohen MS, Dye C, Fraser C, Miller WC, Powers KA, et al. (2012) HIV treatment as prevention. Debate and commentary - will early infection compromise treatment-as-prevention strategies? PLOS Med 9: e1001232. doi:10.1371/journal.pmed.1001232
112. Wawer MJ, Gray RH, Sewankambo NK, Serwadda D, Li X, et al. (2005) Rates of HIV-1 transmission per coital act, by stage of HIV-1 infection, in Rakai, Uganda. J Infect Dis 191: 1403–1409.
113. Gisquet-Quint D, Poterat J (2005) Questioning Wawer et al.’s estimated rate of sexual HIV transmission from persons with early HIV infections. J Infect Dis 192: 1497–1499; author reply 1499–1500.
114. Powers KA, Ghani AC, Miller WC, Hoffman IF, Pettifor AE, et al. (2011) The role of acute and early HIV infection in the spread of HIV and implications for transmission prevention strategies in Lilongwe, Malawi: a modelling study. Lancet 378: 256–268.
115. Anvert B, Males S, Purren A, Taljaard D, Carel M, et al. (2004) Can highly active antiretroviral therapy reduce the spread of HIV?: A study in a township of South Africa. J Acquir Immune Defic Syndr 36: 613–621.
116. Gauldler IF, Watkins DI (2008) Impact of MHC class I diversity on immune control of immunodeficiency virus replication. Nat Rev Immunol 8: 619–630.
117. Seth PM, Danesh A, Sheung A, Rebapragada A, Shahabi K, et al. (2006) Disproportionately high semen shedding of HIV is associated with compartmentalized cytomegalovirus reactivation. J Infect Dis 193: 45–48.
118. Chinek N, Taelman H, Herrmann P, Piot P, Schounmacher M, et al. (1989) A cluster of HIV infection among heterosexual people without apparent risk factors. N Engl J Med 321: 1460–1462.
119. (1999 Mar 5) Williams enters conditional plea of guilty in HIV exposure case. AIDS Policy Law 14: 10.
120. Abu-Raddad LJ, Patnaik P, Kublin JG (2006) Dual infection with HIV and malaria fuels the spread of both diseases in sub-Saharan Africa. Science 314: 1603–1606.
121. Hoffman IF, Jere CS, Taylor TE, Munthali P, Dyer JR, et al. (1999) The effect of Plasmodium falciparum malaria on HIV-1 RNA blood plasma concentration. AIDS 13: 487–494.