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# **Cost-effectiveness of peer- versus venue-based approaches for detecting undiagnosed HIV among heterosexuals in high-risk New York City neighborhoods**

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# **Abstract**

**Introduction—**We used a computer simulation of HIV progression and transmission to evaluate the cost-effectiveness of a scale-up of three strategies to seek out and test individuals with undiagnosed HIV in New York City (NYC).

**Setting—**Hypothetical NYC population

**Methods—**We incorporated the observed effects and costs of the three "seek and test" strategies in a computer simulation of HIV in NYC, comparing a scenario in which the strategies were scaled up with a one-year implementation or a long-term implementation with a counterfactual scenario with no scale-up. The simulation combined a deterministic compartmental model of HIV transmission with a stochastic microsimulation of HIV progression, calibrated to NYC epidemiological data from 2003 to 2015. The three approaches were respondent driven sampling (RDS) with anonymous HIV testing ("RDS-A"), RDS with a two-session confidential HIV testing approach ("RDS-C"), and venue-based sampling ("VBS").

**Results—**RDS-A was the most cost-effective strategy tested. When implemented for only one year and then stopped thereafter, using a societal perspective, the cost per quality-adjusted lifeyear (QALY) gained versus no intervention was \$812/QALY, \$18,110/QALY, and \$20,362/QALY for RDS-A, RDS-C, and VBS, respectively. When interventions were implemented long-term, the cost per QALY gained versus no intervention was cost-saving, \$31,773/QALY, and \$35,148/ QALY for RDS-A, RDS-C, and VBS, respectively. When compared to RDS-A the incremental cost effectiveness ratios (ICERs) for both VBS and RDS-C were dominated.

**Conclusion—**The expansion of the RDS-A strategy would substantially reduce HIV-related deaths and new HIV infections in NYC, and would be either cost-saving or have favorable costeffectiveness.

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### **Keywords**

cost effectiveness; undiagnosed HIV; respondent driven sampling; venue based sampling; heterosexuals; health disparities; HIV testing

# **Introduction**

Individuals with an undiagnosed HIV infection are linked to nearly a third of all HIV transmission events in the United States  $(US)$ .<sup>1</sup> Indeed, the aim of ending HIV transmission in the US cannot be achieved without improved methods to identify these individuals with undiagnosed HIV. In the effort to uncover these undiagnosed cases, the Centers for Disease Control and Prevention (CDC) has recommended persons at high risk for HIV infection receive diagnostic HIV testing at least annually.<sup>2</sup> Yet approximately 13% of persons living with HIV (PLWH) are unaware of their status<sup>3</sup> and identifying these undiagnosed HIV cases, the first step to engagement along the HIV care continuum, remains a significant challenge.<sup>4</sup>

Insufficient HIV diagnostic testing, leading to elevated rates of undiagnosed HIV infection, disproportionately affects heterosexual populations at high-risk for HIV (HHR) compared to other risk groups.<sup>3</sup> Consistent with the CDC National HIV Behavioral Surveillance (NHBS) studies, we define HHR as those who are heterosexually active and socially networked within geographic areas with both excess HIV burden and socioeconomic disadvantage, referred to as "high-risk areas."<sup>5, 6</sup> Overall, less than half of heterosexuals have tested for HIV in their lives (43.5%), compared to more than two-thirds (69%) of men who have sex with men (MSM).<sup>7–9</sup> Among the subpopulation of HHR, most ( $>$  90%) have been tested for HIV at least once in the lifetimes, but regular, annual testing is uncommon.10 This is a concern because HIV prevalence is substantially higher among HHR, who are predominantly African American/Black and Hispanic, than among heterosexuals in the general population  $(2.0\% \text{ vs. } 0.1).$ <sup>11</sup>

Barriers at individual/attitudinal-, social-, and structural-levels of influence impede access to and uptake of regular, annual HIV testing among HHR, including insufficient knowledge of HIV, substance use, fear of stigma, distrust of medical settings, social norms that deter HIV testing, and poor access to settings where high-quality HIV testing is offered.<sup>12, 13</sup> Accordingly, in 2010, the National Institute on Drug Abuse (NIDA) at the National Institutes of Health called for research on new approaches to seek out persons with undiagnosed HIV, provide them with HIV counseling and testing, and link those found to be HIV infected into medical care, with high retention, which are referred to as "Seek, Test, Treat, and Retain" (STTR) studies.14 Under this initiative, we recently evaluated a set of innovative seek and test approaches in a high-risk area (due to elevated rates of heterosexual HIV prevalence and poverty) in the borough of Brooklyn in New York City (NYC). This past study, which focused on African American/Black and Hispanic HHR to provide culturally based intervention approaches, tested three different seek and test strategies designed to be acceptable and easy to access for this high-risk population including: 1) participant recruitment through respondent-driven sampling (RDS) and confidential HIV testing in two

sessions [referred to as RDS-Confidential (RDS-C)]; 2) participant recruitment through RDS and anonymous HIV testing in one session, where the participant's name was not recorded [RDS-Anonymous (RDS-A)]; and 3) participant recruitment through venue-based sampling (VBS) and confidential HIV testing in a single session (Figure 1).<sup>15</sup>

Our previous results<sup>15</sup> demonstrated that the use of RDS, a peer-referral approach, was more efficacious than VBS for uncovering HHR with undiagnosed HIV. We found RDS-A (2.8%, 95% CI 1.1% - 4.3%) and RDS-C (1.0%, 95% CI 1.1% - 1.9%) yielded significantly higher rates of newly diagnosed HIV than VBS (0.5%, 95% CI 0.2% - 0.8%), not accounting for RDS weighting associated with social network size.15 These peer-based strategies, therefore, may have a vital role to play in efforts to eliminate HIV transmission, but they are costly. Therefore, assessing their cost-effectiveness is a necessary prerequisite for considering scale-up. The objective of the present study was to use a computer simulation of HIV progression and transmission to evaluate and compare the cost-effectiveness of a scale-up of these three seek and test strategies in NYC.

# **Methods**

A previously validated simulation of HIV progression and transmission<sup>16, 17</sup> was modified to incorporate the observed effects and costs of the three seek and test interventions. Using the simulation, the impact and cost-effectiveness of a scale-up of the seek and test interventions in NYC were estimated.

#### **Model Overview**

The simulation integrates information from an individual-based stochastic Monte Carlo microsimulation of HIV progression with a deterministic compartmental model of HIV transmission.<sup>16, 17</sup> The simulation is composed of two models. The first model is a natural history model that follows a cohort of HIV-infected patients and predicts time until HIV antiretroviral therapy (ART) failure, accumulation of resistance mutations, and patient survival. Patients progress to AIDS and AIDS-related deaths at varying rates depending on whether they adhere to ART regimens and/or develop resistance to ART, based on HIV viral load suppression and CD4 trajectory. This progression model provides data to inform the second model, a transmission model. The model was developed using C/C++. This process is described in detail in Section 2.1 of the Appendix.

The transmission of HIV through the NYC population is predicted by a compartmental model. Segments of a hypothetical population can become HIV infected, have their infection detected, and access treatment, which can modify their infectivity. Segments of this population can also modify their infectivity by exhibiting risk behaviors including multiple sexual partnerships, neglecting to use condoms, having sexually transmitted infections (STIs), and using illicit substances. The model includes both sexual transmission of HIV and transmission through syringe and injection-related paraphernalia-sharing during injection drug use (IDU). HIV transmission was modeled using a binomial process and assumed assortative mixing in the population. The probability of transmission between partners was adjusted to account for infected partner's gender, disease state, and treatment status.

Transmission model compartments are stratified by age, sexual activity level, presence of unhealthy alcohol use, IDU, HIV status, and if infected, HIV viral load, CD4, and ART resistance patterns. The design of the simulation, as well as its calibration and validation, is described in more detail in the Appendix and elsewhere.<sup>16, 17</sup> We used the calibrated simulation to evaluate the impact and value of the three seek and test interventions in NYC.

The seek and test interventions' effects were represented in the transmission model pathway by accelerating transitions from "infected but undetected" compartments to "infected and detected" compartments. We conservatively assumed that intervention effects only persisted while the respective intervention program was continued. The simulation was calibrated to NYC epidemiological data with the goal of replicating trends in NYC HIV prevalence, incidence, deaths, and persons with HIV from 2003 to 2015. The analyses performed were assumed to have started in 2015 with an estimated 2015 HIV prevalence, deaths, and incidence being represented in the model. The model inputs and the intervention effects are described in more detail in Table 1.

To examine hypothetical effects on HIV with scaling of the seek and test programs, we varied "reach," which we defined as the proportion of the population with an unknown HIV status enrolled in the program. To compare the costs and impact of each intervention for a hypothetical implementation of just one year, the reach of the interventions was set to 27.5 persons enrolled per 1,000 population, correlating to the level at which the most effective intervention (RDS-A) identified 50% of the previously unknown HIV-positive population in one year. A one year implementation was selected to investigate the impact of a potential programmatic choice for a one-off short term implementation. To evaluate the costs and impact of the intervention implemented long-term over a 20-year horizon the program reach was set to 4.5 persons enrolled per 1,000 population, correlating to the level at which the most effective intervention (RDS-A) identified 50% of the previously unknown HIV-positive population after 20 years. The benchmark of 50% was chosen arbitrarily.

Costs and effects were discounted at 3%, our time horizon was 20 years, and costs were assessed from a societal perspective, which included the cost of treatment, as well as, a programmatic perspective, that did not include treatment costs, using 2015 \$US. Other than specifying a finite time horizon, all other aspects of the cost-effectiveness analysis were conducted in line with recommendations by the Panel on Cost-Effectiveness in Health and Medicine.18 We chose a 20 year rather than infinite time horizon because we found that it was the longest time horizon viewed as credible by stakeholders (e.g. public health decision makers) in general. The simulated population consisted of HIV-infected and uninfected New Yorkers from 2015 through 2035, with the intervention assumed to begin in the year 2015. More complete details of model specification, initial population structure and parameterization are located in the appendix and have been published elsewhere.<sup>16, 17</sup>

#### **Seek and Test Interventions**

As noted above, the seek and test study included three different interventions designed to be acceptable and easy to access for this high-risk population including: 1) respondent driven sampling (RDS) with anonymous HIV testing ("RDS-A"), 2) RDS with a two-session confidential HIV testing approach ("RDS-C"), 3) and venue-based sampling ("VBS")

(Figure 1).15 All three seek and test interventions included components to refer those testing HIV-negative to prevention services, and link newly diagnosed individuals to HIV primary care (i.e., a "treat & retain" study phase). Complete methods are described elsewhere.<sup>19–21</sup> The primary outcome that was incorporated as an input into the HIV model was the number of newly diagnosed HIV infections per number enrolled ("hit rate") and the cost per patient of each intervention (Table 1).

#### **Cost-effectiveness analysis**

We conducted simulations where the seek and test interventions were activated, and calculated the health benefits, costs, and cost-effectiveness ratios of each over the 20-year time horizon. These simulations were compared to a reference case where no additional interventions were implemented. Outcomes measured include total quality-adjusted life years (QALYs) gained, cost per QALY gained, incremental cost effectiveness ratios (ICERs), number and proportion of new HIV infections averted, incremental cost per infection averted, infections detected and cost per infection detected. As a sensitivity analysis, we varied both intervention efficacy and cost independently across plausible ranges and evaluated their impact on cost-effectiveness. We then simulated the seek and test interventions and calculated the ICERs of the interventions. ICERs measure the additive benefit of each strategy compared with its next best alternative, and interpret this benefit together with its additive cost. A cost per QALY gained value less than \$100k was considered cost-effective because it approximates the opportunity cost of achieving health benefit in the US health system and a cost per QALY gained value less than \$20k was considered very cost effective because it approximates the opportunity cost of highly effective HIV-specific ART.<sup>22</sup>

## **Results**

#### **Effectiveness**

**One-Year Implementation—**When implemented for only one year and then stopped thereafter, the RDS-A, RDS-C, and VBS interventions reduced the number of new HIV infections over 20 years by 533, 126, and 64 infections, respectively, versus the base case of 58,402 new infections in NYC (Figure 2a,b). With the RDS-A, RDS-C, and VBS interventions the number of QALYs gained was 17,191, 4,288, and 2,206, respectively (Figure 2c). With the RDS-A, RDS-C, and VBS interventions the number of HIV-related deaths over 20 years was reduced by 1,626, 404, and 208, respectively versus a base case of 14,324 deaths.

**Long-term Implementation—**When implemented long-term, the RDS-A, RDS-C, and VBS interventions reduced the number of new HIV infections over 20 years by 2,166, 657, and 344 infections, respectively, versus base case of 58,402 HIV-infected persons in NYC (Figure 3a,b). With the RDS-A, RDS-C, and VBS interventions the number of QALYs gained was 15,291, 5,040, and 2,683, respectively (Figure 3c). With the RDS-A, RDS-C, and VBS interventions the number of HIV-related deaths over 20 years was reduced by 2,769, 916, and 488, respectively versus a base case of 14,324 deaths.

#### **Cost effectiveness**

The cost per person enrolled in each strategy was \$531, \$648, and \$368 for RDS-A, RDS-C, and VBS, respectively. From a programmatic perspective, the cost per newly detected HIVpositive person was \$18,882, \$69,612, and \$74,206 for RDS-A, RDS-C, and VBS, respectively.

**One-Year Implementation—**When implemented for only one year and then stopped thereafter, and using a programmatic perspective, RDS-A, RDS-C, and VBS added discounted costs of \$74,556,402, \$91,019,837, and \$51,693,795, respectively; corresponding to costs per infection averted of \$139,808, \$724,678, and \$805,058, respectively. The cost per QALY gained versus no intervention was \$4,337/QALY, \$21,225/QALY, and \$23,431/ QALY for RDS-A, RDS-C, and VBS, respectively. When the three alternatives were considered together, the ICER for RDS-A was \$1,526 compared to VBS, and RDS-C was dominated. (Table 2)

When implemented for only one year and then stopped thereafter, but using a societal perspective, over 20 years the one-year implementation of RDS-A, RDS-C, and VBS added discounted costs of \$44,684,670,725, \$44,748,362,898, and \$44,715,626,205, respectively; corresponding to costs per infection averted of \$26,191, \$61,8306, and \$699,608, respectively. The cost per QALY gained versus no intervention was \$812/QALY, \$18,110/ QALY, and \$20,362/QALY for RDS-A, RDS-C, and VBS, respectively. When the three alternatives were considered together, compared to RDS-A the ICERs for both VBS and RDS-C were dominated. (Table 2)

**Long-term Implementation—**When implemented long-term but using a programmatic perspective RDS-A, RDS-C, and VBS resulted in a total discounted cost of \$195,522,868, \$238,661,689, and \$135,543,430, respectively; corresponding to costs per infection of \$90,275, \$363,252, and \$393,662, respectively. The cost per QALY gained versus no intervention was \$12,787/QALY, \$47,356/QALY, \$50,523/QALY for RDS-A, RDS-C, and VBS, respectively. When the three alternatives were considered together, the ICER for RDS-A was \$4,757 compared to VBS, and RDS-C was dominated. (Table 2)

When implemented long-term but using a societal perspective, RDS-A, RDS-C, and VBS resulted in a total discounted cost of \$44,610,518,121, \$44,830,830,915, and \$44,764,998,002, respectively; corresponding to costs per infection averted of cost-saving, \$243,720, and \$273,862, respectively. The cost per QALY gained versus no intervention was cost-saving, \$31,773/QALY, and \$35,148/QALY for RDS-A, RDS-C, and VBS, respectively. When compared to RDS-A the ICERs for both VBS and RDS-C were dominated. (Table 2)

#### **Sensitivity Analyses**

In sensitivity analyses of both a one-year and long-term implementation, all three interventions maintained favorable value across all costs and "hit rate" values assessed (Figure 4). In a one-year implementation, for the RDS-A strategy to become cost-saving, the cost of the program must be decreased by approximately 22.6% from \$530 to \$410. In long-

term implementation, for the RDS-A strategy to remain cost-saving the strategy's "hit rate" must be decreased by no more than 28.6% from 0.028 to 0.02 nor can the cost be increased by more than 37.7% from \$530 to \$730. To be cost-saving the RDS-C and VBS strategies would require significant reductions in programmatic costs. (Figure 4)

# **Conclusion**

We provide estimates of the impact and cost-effectiveness of the scale-up of three seek and test strategies for HHR conducted in a large urban high-risk area. Our analyses suggested that, from a societal perspective, a scale-up of RDS-A using a long-term implementation was likely to be cost-saving and dominated the two other seek and test strategies. Even when the strategies were limited to one year and discontinued thereafter, RDS-A remained costeffective and dominated the other strategies investigated. Our findings were robust over a range of assumptions regarding cost and effectiveness. While other RDS-based strategies have been used in numerous research studies and intervention delivery settings for other high-risk populations, such as MSM and persons who inject drugs,  $^{23}$  and the CDC's NHBS projects have used RDS for recruiting HHR populations nationally,  $24$ ,  $25$  RDS and VBS seek and test strategies for HIV testing in HHR have not previously been evaluated for costeffectiveness.

Our findings suggest that RDS-A, a single-session and low-threshold approach designed to be acceptable and easy to access for this high-risk population, is a cost-effective seek and test strategy for identifying HIV-infected persons who were otherwise undetected, overcoming many of the challenges to engaging HHR in HIV testing. In fact, past research has shown high-quality HIV testing experiences may foster more timely engagement in HIV care, in contrast to negative, or even coercive, experiences that may impede acceptance of the new HIV status and trigger medical distrust and fear.<sup>26</sup> Descriptive exploratory findings from the "treat and retain" phase that followed the study's seek and test phase, albeit with imprecise estimates given small sample sizes, showed majority of those newly diagnosed with HIV in RDS-A elected to enroll in a confidential treat and retain phase designed to foster linkage to HIV care in a timely fashion, and of these, the majority (> 80%) had achieved undetectable HIV viral load levels within six months.<sup>27</sup> Similarly, the majority of participants found newly diagnosed with HIV in RDS-C engaged in HIV care and achieved undetectable HIV viral load levels ( $> 60\%$ ) by the final follow-up period, approximately a year after diagnosis.28 In sum, our cost-effectiveness analyses provide the strongest support for the implementation and scale-up of a RDS-A seek and test strategy in NYC and, we estimate, comparable urban environments, followed by RDS-C.

While all three seek and test strategies investigated were cost-effective compared to no intervention, the RDS-A strategy proved to be the most cost-effective RDS strategy tested, because not only was it more effective at finding undiagnosed cases of HIV in HHR,<sup>21</sup> but it also incurred a lower programmatic cost than RDS-C. Further, while both RDS strategies performed better than VBS at identifying individuals with high rates of serious risk factors such as incarceration, unemployment, and homelessness, the RDS-A strategy outperformed RDS-C in this regard. We speculate that RDS-A out-performed RDS-C in part because it successfully engaged participants with high rates of multiple risk factors, including those

with current or past substance use problems, who typically experience serious barriers to HIV testing and services. This, combined with the provision of HIV testing at the first contact, made the RDS-A intervention easy to access, and also, we speculate, helped reduce the fear of potential HIV stigma through anonymous testing.<sup>21</sup> The same mechanism that may have reduced perceived stigma among participants; namely, anonymous testing in the first session, also acted to reduce the required programmatic investment per person enrolled. By decreasing personnel resources consumed by providing testing in a single session, the overall programmatic costs of RDS-A-based testing were reduced by roughly 18%. Thus single-session/anonymous strategy peer-referral approaches for uncovering HHR with undiagnosed HIV can reduce costs and have a vital role to play in efforts to eliminate HIV transmission.

In scenarios in which seek and test implementation lasted for only one year and was discontinued thereafter, the number of infections averted peaked approximately 8 years after program implementation for all three strategies. This finding can be attributed to the generation of new infections by the patients on ART that would have otherwise not occurred due to death from AIDS without the intervention. Although treatment reduces the infectivity of an individual, it does not completely eliminate the risk of HIV transmission;<sup>29</sup> consequently, on treatment these patients are living longer and have a greater number of opportunities to spread HIV. In the base case scenario, however, these individuals do not have the long-term potential to infect others due to the high probability of death.

Similarly, the RDS-A strategy can be seen to accrue a greater number of QALYs gained in a one year implementation as compared to a long-term intervention. The greater number of QALYs gained is a result of placing a greater number of people on ART at an earlier time, therefore allowing those on treatment to generate an increased rate of QALYs for longer. The difference between the long-term implementation and the one-year implementation, however, decreases with time as more individuals are placed on treatment in the long-term implementation, and less effective interventions such as RDS-C and VBS, which have a greater number of QALYs gained at shorter time horizons, can be seen to have fewer QALYs gained at 20 years than a long-term implementation.

The waning impact of a single wave of seek and test emphasizes the need for a long-term implementation. By achieving cost-saving at 5-years, scaled up implementation of RDS-A has the potential to have great health impact (prevent 0.9% of infections and 11.4% of deaths) while being cost-saving in the short-term. Long-term implementation, while not necessarily cost-saving in the short-term, are cost-effective, have the potential to prevent 3.7% of infections and 19.3% of deaths, and are cost-saving in the long term.

Our analysis has a number of limitations. First, our model did not incorporate the potential limitations of RDS as a sampling strategy, including potentially diminishing returns as a smaller proportion of HIV-positive cases remain undetected. This may underestimate the cost per infection identified for the two RDS strategies, as over time more people may need to be enrolled in order to identify a consistent number of infections. However, we limited our scale-up assumption to 50% to minimize the bias of diminishing marginal returns. Second, the model cannot distinguish between the identification of new infections and HIV-positive

individuals that have been previously tested and diagnosed. Those previously diagnosed but mislabeled as newly diagnosed may lead to overestimation of cost-effectiveness, as some will have made progress on the continuum of care prior to the program. Also, those previously diagnosed but not linked to care, are more likely to have characteristics that are associated with being less likely to receive ART and stay engaged in care than a newly diagnosed individual.<sup>30, 31</sup> Third, the analysis included the same intervention costs for all individuals enrolled in the study, including those that did not ultimately receive HIV testing. This, however, likely overestimates the total cost of implementation as individuals not receiving HIV testing would not incur the full intervention cost. Fourth, this study was unable to distinguish whether different populations were reached by the different strategies. Therefore, further study may be needed to determine whether unique populations are reached by each strategy and to assess whether combination strategies may be appropriate. Finally, the seek and test interventions were carried out in a population that has a relatively high prevalence of HIV and HHR compared to other populations in the US. Consequently our results may not be applicable to all populations. However, the HIV prevalence rate is similar to high-risk neighborhoods in 11 of the 15 most populous cities in the US, as well as, in numerous other smaller cities.<sup>32, 33</sup> Further testing is needed to investigate the RDS-A strategy's effectiveness in populations outside of densely populated high risk urban areas.

The expansion of the RDS-A seek and test strategy would substantially reduce HIV-related deaths and avert new HIV infections in NYC, and would be either cost-saving or have favorable cost-effectiveness. Further testing is needed to investigate the strategy's effectiveness in urban populations outside of NYC and outside of densely populated high risk urban areas.

# **Supplementary Material**

Refer to Web version on PubMed Central for supplementary material.

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**Figure 1. Schematic of study activities**

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#### **Figure 2. Impact of a 1 year implementation (a) infections averted, (b) new infections, (c) QALYS gained**

Note: The decrease in infections averted at the 20-year time horizon can be attributed to the generation of new infections by the patients on treatment that are living longer and have a greater number of opportunities to spread HIV. In the base case scenario, however, these individuals do not have the long term potential to infect others due to the high probability of death.

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**Figure 3. Impact of a long-term implementation (a) infections averted, (b) new infections, (c) QALYS gained**

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| A   |           | Cost of intervention per person enrolled |           |          |                        |          |          |          |                            |          |          |                          |          |          |  |
|---|-----------|--|-----------|----------|------------------------|----------|----------|----------|----------------------------|----------|----------|--------------------------|----------|----------|--|
|   | \$210     | \$250                                    | \$290     | \$330    | \$370                  | \$410    | \$450    | \$490    | \$530                      | \$570    | \$610    | \$650                    | \$690    | \$730    |  |
| 0.005                                     | \$10,190  | \$12,716                                 | \$15,242  | \$17,768 | <b>VBS</b><br>\$20,293 | \$22,819 | \$25,345 | \$27,870 | \$30,396                   | \$32,922 | \$35,447 | \$37,973                 | \$40,499 | \$43,024 |  |
| 0.01                                      | \$3,242   | \$4,454                                  | \$5,667   | \$6,879  | \$8,092                | \$9,304  | \$10,517 | \$11,729 | \$12,942                   | \$14,154 | \$15,367 | <b>RDS C</b><br>\$16,579 | \$17,792 | \$19,004 |  |
| $\frac{1}{2}$ 0.015<br>$\frac{1}{2}$ 0.02 | \$854     | \$1,624                                  | \$2,394   | \$3,164  | \$3,935                | \$4,705  | \$5,475  | \$6,245  | \$7,015                    | \$7,785  | \$8,555  | \$9,326                  | \$10,096 | \$10,866 |  |
|   | $-5420$   | \$124                                    | \$667     | \$1,211  | \$1,755                | \$2,298  | \$2,842  | \$3,385  | \$3,929                    | \$4,473  | \$5,016  | \$5,560                  | \$6,104  | \$6,647  |  |
| 0.025                                     | $-$1,299$ | $-$ \$899                                | $-5500$   | $-5100$  | \$300                  | \$699    | \$1,099  | \$1,498  | \$1,898                    | \$2,298  | \$2,697  | \$3,097                  | \$3,496  | \$3,896  |  |
| 0.028                                     | $-51,791$ | $-51.462$                                | $-51.133$ | $ $804$  | $-$475$                | $-5146$  | \$183    | \$512    | <b>RDSA</b><br><b>S841</b> | \$1,170  | \$1,499  | \$1,828                  | \$2,158  | \$2,487  |  |



#### **Figure 4. Sensitivity analyses for (a) one-year implementation and (b) long-term implementation from a societal perspective**

Note: "Hite rate" represents number of newly diagnosed HIV infections per number enrolled; VBS = venue-based sampling; RDS-C = respondent driven sampling, confidential; RDS-A = respondent driven sampling, anonymous; QALY = quality adjusted life year

#### **Table 1**

## Key Input parameters





ART: antiretroviral therapy

# **Table 2**

 $\mathbf{I}$ 

Twenty Year Time Horizon Intervention Incremental Cost-Effectiveness Twenty Year Time Horizon Intervention Incremental Cost-Effectiveness

