Incidence and Prevalence of Intrasubtype HIV-1 Dual Infection in At-Risk Men in the United States

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Background. Human immunodeficiency virus type 1 (HIV-1) dual infection (DI) has been associated with decreased CD4 T-cell counts and increased viral loads; however, the frequency of intrasubtype DI is poorly understood. We used ultradepth sequencing (UDS) to estimate the frequency of DI in a primary infection cohort of predominantly men who have sex with men (MSM).

Methods. HIV-1 genomes from longitudinal blood samples of recently infected, therapy-naive participants were interrogated with UDS. DI was confirmed when maximum sequence divergence was excessive and supported by phylogenetic analysis. Coinfection was defined as DI at baseline; superinfection was monoinfection at baseline and DI at a later time point.

Results. Of 118 participants, 7 were coinfected and 10 acquired superinfection. Superinfection incidence rate was 4.96 per 100 person-years (95% confidence interval [CI], 2.67–9.22); 6 occurred in the first year and 4 in the second. Overall cumulative prevalence of intrasubtype B DI was 14.4% (95% CI, 8.6–22.1%). Primary HIV-1 incidence was 4.37 per 100 person-years (95% CI, 3.56–5.36).

Conclusions. Intrasubtype DI was frequent and comparable to primary infection rates among MSM in San Diego; however, superinfection rates declined over time. DI is likely an important component of the HIV epidemic dynamics, and development of stronger immune responses to the initial infection may protect from superinfection.

Keywords. intrasubtype HIV-1 dual infection; HIV-1 coinfection; HIV-1 superinfection; ultradepth sequencing; acute and early infection.

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next-generation massively parallel sequencing modalities (eg, ultradeep pyrosequencing [UDS]) can identify DI more accurately and efficiently than previous methods [16]. Multiple platforms of UDS can detect circulating minority variants as low as 0.25% of the viral population [17, 18]. Optimization of high-throughput UDS protocols can further increase the efficacy of screening for evidence of DI. Using UDS, one group recently reported rates of superinfection comparable to those of primary HIV-1 infection in a heterosexual cohort in Rakai, Uganda [19]. Our group and others have also reported associations between DI and increased viral load [7, 10, 15, 20], faster decline in CD4+ T cells, and shorter time to AIDS diagnosis [21, 22]. A more accurate estimate of the frequency of DI may not only help paint a clearer picture of viral transmission dynamics within human populations but also assess the potential impact of DI on disease progression and correlates of protection from superinfection. To this end, we employed UDS to screen longitudinally collected blood plasma samples from recently infected individuals to determine the rates of coinfection and superinfection in San Diego, California.

**METHODS**

**Study Participants and Clinical Data**

This study included participants enrolled in the San Diego Primary Infection Cohort between January 1998 and January 2007 who were antiretroviral therapy (ART) naive, had deferred ART for at least 6 months after their estimated date of infection (EDI), and had at least 2 blood plasma samples available for sequencing. EDI was calculated at baseline for each participant using a well-established algorithm that characterizes stages of infection on the basis of serologic and virologic criteria [23]. Study participants were seen at weeks 2, 4, 8, 12, 16, 24, 48, and every 24 weeks thereafter for the duration of the study; and those who started ART as part of routine clinical care were followed every 8 weeks during ART for 24 weeks or until plasma viral load (VL) was ≤50 copies/mL. Upon collection, samples were aliquoted, frozen, and stored at −80°C. At all time points, CD4 T-cell counts (LabCorp), and blood plasma HIV-1 RNA levels (Amplicor HIV-1 Monitor Test, Roche Molecular Systems Inc.) were quantified. HIV-1 subtype was determined for each subject using HIV-1 pol sequence data generated by Viroseq 2.0 (Applied Biosystems) by the SCUEAL algorithm [24]. Demographics, clinical data, and standard laboratory values were also collected. Behavioral risk factors were assessed at baseline and longitudinally for all participants, as described elsewhere [25]. Similar assessments were conducted for individuals receiving HIV screening in the Early Test program, which includes antibody and nucleic acid testing [25].

**RNA Extraction and Sequencing Methods**

HIV-1 RNA was isolated from cryostored blood plasma (QIAamp Viral RNA Mini Kit, Qiagen, Hilden, Germany), and complementary DNA (cDNA) generated (RETROscript Kit, Applied Biosystems/Ambion, Austin, TX) according to manufacturers’ instructions. As described previously [16, 20], 500 μL or 1000 μL of blood plasma were used for samples with viral loads above or below 20,000 copies/mL, respectively. Three coding regions—gag p24 (HXB2 coordinates 1366–1619), pol reverse transcriptase (RT, HXB2 coordinates 2708–3242), and env C2–V3 (HXB2 coordinates 6928–7344)—were amplified by polymerase chain reaction (PCR) with region-specific primers, and rubber gaskets were used to physically separate 16 concurrently sequenced samples on a single 454 GS FLX titanium picoliter plate (454 Life Sciences/Roche, Branford, CT) [16, 20, 26]. To avoid cross-contamination, extraction procedures, PCR and UDS library preparation of samples from the same cohort participants were performed on different days and in different sequencing runs. UDS data complemented prior single genome sequencing (SGS) data of env C2–V3 and pol RT generated for 49 cohort participants and published elsewhere [16, 20]. For each sample, the cDNA template input was calculated assuming 43% reverse transcription efficiency and was expressed as the number of templates (log_{10}) present in the 10 μL reaction volume used for the first round of nested PCR (ie, cDNA input before any PCR amplification procedure). Average reverse transcription efficiency was validated by quantifying the input cDNA for 19 random samples from the total pool with realtime quantitative PCR (qPCR) and comparing these results with the HIV RNA measurements, as described elsewhere [26].

**Sequence Analysis and Dual Infection Confirmation**

UDS data were analyzed using the 454 UDS bioinformatics pipeline, available as part of the HyPhy software package [27] and the Datamonkey web server [24]. The methodology has been described in detail elsewhere [20]. Briefly, high-quality reads were retained and aligned to HXB2 library reference genes for the 3 coding regions using an iterative codon-based alignment procedure, and the maximum divergence (MDI) was calculated for each coding region as the maximum likelihood divergence using the HKY85 substitution model in sliding windows of 150 nucleotides (50-nucleotide shifts, minimum site coverage 500 reads). The output alignments of all UDS datasets have been uploaded to the Los Alamos National Laboratory HIV-DB Next Generation Sequence Archive (http://www.hiv.lanl.gov/content/sequence/HIV/NextGenArchive) and are also available upon request from the corresponding author. Putative DI was suggested when the MDI was >5% in env and >2.5% in gag and pol. If the MDI exceeded these thresholds in at least one coding region, standard phylogenetic analyses (neighbor-joining and maximum likelihood tree reconstruction and bootstrapping) were performed and highlighter plots were created using 500 randomly selected reads, and if 2 clusters were separated by a branch with a bootstrap value >90%, the sample was preliminarily classified as DI. Representative
RESULTS

At study baseline, participants (n = 118) were infected with HIV-1 subtype B and were predominantly white men (82%) with a median age of 32 years (IQR: 26–38 years) who reported sex with other men as their main HIV risk factor (Table 1). At enrollment, the median CD4⁺ T-cell count was 544 cells/mL (IQR: 460–709 cells/mL), the median time from EDI was 71 days (IQR: 70–133 days), and the median blood plasma level of HIV-1 RNA was 4.81 log₁₀ copies/mL (IQR: 3.96–5.38 log₁₀ copies/mL). Further breakdown of CD4⁺ T-cell count, plasma HIV-1 RNA, and other clinical markers according to dual infection status is outlined in Table 1. The median time from EDI to ART initiation in the group was 16.2 months (IQR: 6.7–33.2 months). When using UDS to screen for DI, the median calculated cDNA input in the first-round nested PCR was 3.42 log₁₀ copies/10 μL (IQR: 3.13–3.90 log₁₀ copies); the median number of UDS reads generated passing quality filters for partial env was 3877 (IQR: 1771–6417) with a median nucleotide coverage of 2032 (IQR: 944–3492), the median number of reads generated for partial gag was 8263 (IQR: 3189–12 911) with a median coverage of 3018 (IQR: 1093–4927), and for partial pol, the median read number was 4268 (IQR: 1911–8082) with a median per site nucleotide coverage of 1320 (IQR: 608–2619; Table 2).

At baseline, 7 cases of coinfection were identified at a median time from EDI of 2.8 months (IQR: 2.3–3.2 months after EDI), and a coinfection prevalence of 5.9% (95% CI, 2.4%–11.8%). During follow-up, 10 participants were identified with superinfection over 201.6 person-years of follow-up. This

### Table 1. Study Cohort Baseline Characteristics

<table>
<thead>
<tr>
<th>Factor</th>
<th>Monoinfection (N = 101)</th>
<th>Coinfection (N = 7)</th>
<th>Superinfection (N = 10)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age at enrollment: median years (IQR)</td>
<td>31 (27–39)</td>
<td>34 (27–39)</td>
<td>31 (24–34)</td>
</tr>
<tr>
<td>Sex, no. (%)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Male</td>
<td>98 (97)</td>
<td>7 (100)</td>
<td>10 (100)</td>
</tr>
<tr>
<td>Female</td>
<td>3 (3)</td>
<td>0 (0)</td>
<td>0 (0)</td>
</tr>
<tr>
<td>Race, no. (%)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>White</td>
<td>84 (83)</td>
<td>4 (57)</td>
<td>8 (80)</td>
</tr>
<tr>
<td>Black</td>
<td>7 (7)</td>
<td>0 (0)</td>
<td>1 (10)</td>
</tr>
<tr>
<td>Native American</td>
<td>6 (6)</td>
<td>0 (0)</td>
<td>1 (10)</td>
</tr>
<tr>
<td>Hawaiian or Pacific Islander</td>
<td>1 (1)</td>
<td>2 (29)</td>
<td>0 (0)</td>
</tr>
<tr>
<td>Asian</td>
<td>1 (1)</td>
<td>0 (0)</td>
<td>0 (0)</td>
</tr>
<tr>
<td>Unknown</td>
<td>2 (2)</td>
<td>1 (14)</td>
<td>0 (0)</td>
</tr>
<tr>
<td>Ethnicity, no. (%)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Hispanic or Latino</td>
<td>17 (17)</td>
<td>2 (29)</td>
<td>2 (20)</td>
</tr>
<tr>
<td>Not Hispanic or Latino</td>
<td>32 (32)</td>
<td>3 (43)</td>
<td>4 (40)</td>
</tr>
<tr>
<td>Unknown</td>
<td>52 (51)</td>
<td>2 (29)</td>
<td>4 (40)</td>
</tr>
<tr>
<td>HIV risk factors, no. (%)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MSM</td>
<td>98 (97)</td>
<td>7 (100)</td>
<td>10 (100)</td>
</tr>
<tr>
<td>Injection drug use</td>
<td>4 (4)</td>
<td>0 (0)</td>
<td>0 (0)</td>
</tr>
<tr>
<td>HIV viral load: median log₁₀ copies/mL (IQR)</td>
<td>4.84 (4.13–5.38)</td>
<td>4.58 (4.21–5.11)</td>
<td>4.68 (3.16–4.95)</td>
</tr>
<tr>
<td>CD4 T-cell count: median cells/μL (IQR)</td>
<td>542 (460–706)</td>
<td>546 (421–706)</td>
<td>543 (476–685)</td>
</tr>
<tr>
<td>cDNA input: median log₁₀ copies/mL (IQR)</td>
<td>3.51 (3.02–4.04)</td>
<td>3.24 (3.02–3.77)</td>
<td>3.34 (2.12–3.51)</td>
</tr>
</tbody>
</table>

**Abbreviations:** cDNA, complementary DNA; HIV, human immunodeficiency virus; IQR, interquartile range; MSM, men who have sex with men.
resulted in an overall superinfection incidence rate of 4.96 per 100 person-years (95% CI, 2.67–9.22). Six superinfection cases occurred in the first year after EDI, 4 within the first 6 months (first-year incidence rate of 29.20 per 100 person-years, 95% CI, 1.13–6.50), and 4 within the second year (second-year incidence rate of 9.25 per 100 person-years, 95% CI, 3.47–24.64; Figure 1). The median time to superinfection was 10.5 months after EDI (IQR: 5.2–15.6 months). Together, the cumulative prevalence of dual infection (including both coinfections and superinfections) was 14.4% (95% CI, 8.6%–22.1%). All DI cases were infected with 2 subtype B viruses. Phylogenetic trees built using PHYML from longitudinal sequences chosen to represent HIV-1 monoinfection, coinfection, and superinfection cases are shown in Figure 2. Sampling of time points after estimated date of infection for all cohort participants is also listed (Supplementary Table 1).

HIV prevalence in the Early Test program was 2.5% (95% CI, 2.0%–3.1%) among first time testers [25]; and among repeat clients, 92 seroconverted over 21 065 person-years, resulting in an HIV incidence of 4.37 per 100 person-years (95% CI, 3.56–5.36). Behavioral risk factors, including acquisition of sexually transmitted infections, were not significantly different among the MSM identified with primary infection, or DI when assessed longitudinally, although the presence of an STI infection was based on self-report (data not shown).

**DISCUSSION**

Our study used UDS to investigate the incidence and prevalence of intrasubtype DI in a predominantly high-risk cohort of individuals enrolled during primary infection. All identified cases of DI were intrasubtype B, and the cumulative prevalence of DI (both coinfections and superinfections) was high at 14.4%. Of the 17 total cases of DI, 10 superinfections were observed over 215 person-years, giving an incidence of HIV-1 superinfection of 4.96 per 100 person-years. The majority of these superinfection cases occurred in the first year after initial infection, consistent with our group’s prior report [15]. On the other hand, the incidence of primary HIV in our screening program was 4.37 per 100 person-years (among clients who had repeat testing). Although the incidence rates of primary infection and superinfection were not directly compared, they seemed similar. This is consistent with recent reports of the incidence of superinfection in a cohort with different HIV risk factors [19]. The rate of superinfection declined from 29.20 per 100 person-years in the first year to 9.25 per 100 person-years in the second year, which might suggest some protection against superinfection over time; the small number of cases and shorter follow-up from access to ART may bias this observation.

Using a more rigorous disease progression analysis, our group previously reported that HIV-1 superinfection was associated with more rapid rates of viral load increase over time but not with the rates CD4 T-cell count decline [20]. As more cases of DI are identified, more specific immunologic, clinical, and virologic correlates of DI can be determined. One limitation of this study is that while UDS is very sensitive at detecting DI, it is subject to experimental and reproducibility biases, especially when preceded by PCR amplification [28]. Despite increased sensitivity to detect DI in a blood sample, the absence of detected DI does not mean that it did not occur; it could still have been missed by not sampling at the right time, viral recombination, or lack of sensitivity of the UDS platform. The classification of DI in the present study was rigorous. Confirmation of DI at a sampled time point required meeting a series of stringent criteria: excessive sequence divergence, phylogenetic trees consistent with DI, and maintenance of the same phylogenetic relationships in the resampled time point. Although this helped control for sampling biases by the sequencing method and bioinformatics alignment artifacts, the DI call was conservative.

### Table 2. Characteristics of UDS Reads Generated for Each HIV-1 Coding Region

<table>
<thead>
<tr>
<th>Coding Region</th>
<th>No. of Reads: Mean (IQR)</th>
<th>Read Length: Mean (IQR)</th>
<th>Nucleotide Coverage: Median (IQR)</th>
</tr>
</thead>
<tbody>
<tr>
<td>env (C2–V3)</td>
<td>3877 (1771–6417)</td>
<td>296 (269–329)</td>
<td>2032 (944–3492)</td>
</tr>
<tr>
<td>gag (p24)</td>
<td>8263 (3189–12 911)</td>
<td>233 (226–244)</td>
<td>3018 (1093–4927)</td>
</tr>
<tr>
<td>pol (RT)</td>
<td>4268 (1911–8082)</td>
<td>223 (194–242)</td>
<td>1320 (603–2619)</td>
</tr>
</tbody>
</table>

Abbreviations: HIV-1, human immunodeficiency virus type 1; IQR, interquartile range; RT, reverse transcriptase; UDS, ultradepth sequencing.
Figure 2. Phylogenetic trees using PHYML were generated from longitudinal env sequences for participants I4, U1, and L7 illustrating examples of HIV-1 (A) monoinfection, (B) co-infection, and (C) superinfection, respectively. Solid triangles and circles represent haplotypes from different viral strains, open diamonds indicate background sequences, and brackets delineate different timepoints (T1: timepoint 1; T2: timepoint 2; T3: timepoint 3; T4: timepoint 4). The same set of background sequences were used in all trees.
and some DI cases may have been misclassified as monoinfections.

The initiation of ART in the study cohort has the potential to modulate some of the conclusions. First, it limited our ability to assess the incidence of superinfection over longer durations of follow-up. Second, because only ART-naive individuals were considered in the calculation of the incidence of superinfection, this could lead to a selection bias if those who were more (or less) susceptible to superinfection were more (or less) likely to initiate ART. For example, if those who had riskier sex behaviors chose to start ART sooner than those who remained treatment-naive, then our sample would have been biased to consist of a group which was less likely to become superinfected. The opposite scenario could likewise be true, and other unmeasured confounders could have influenced one’s decision to start or not start ART, which would result in unmeasured bias.

Another important caveat is that the group screened for evidence of DI did not comprise HIV-negative individuals wherein at-risk individuals could be followed for estimates of incidence of initial infection. Therefore, a direct comparison of the incidence of primary infection to superinfection in these groups would not be appropriate because the denominators (ie, populations at risk) were different. Because the risks for superinfection are likely the same as those for primary infection, the superinfection group is likely riskier because the person time is for those with one infection already, whereas the person time for initial HIV infection is among all those at risk for infection. Despite these limitations, however, the improvement of our understanding regarding the incidence of intrasubtype HIV superinfection may provide important insight into the correlates of protection necessary to prevent against initial HIV infection, which could inform preventative vaccine design.

**Supplementary Data**

Supplementary materials are available at The Journal of Infectious Diseases online (http://jid.oxfordjournals.org/). Supplementary materials consist of data provided by the author that are published to benefit the reader. The posted materials are not copyrighted. The contents of all supplementary data are the sole responsibility of the authors. Questions or messages regarding errors should be addressed to the author.

**Notes**

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**Author Contributions.** G. A. W. participated in the study design, performed the laboratory experiments, participated in the data analyses for this study and wrote the primary version of the manuscript, M. E. P. participated in the study design, performed the laboratory experiments, participated in the data analyses for this study, and assisted in the writing of the article. G. C. performed the laboratory experiments, S. J. L. designed and implemented the protocol of the study cohort, and assisted in the writing of the article, S. K. P. designed and implemented the bioinformatics analysis pipeline, and assisted in the writing of the article. A. C. participated in data organization and analysis, A. E. R. participated in data analysis and in the editing of the article, S. R. M. participated in data analysis and assisted in the writing of the article, D. D. R. assisted in the writing of the article, D. M. S. conceived the study design, assisted in the collection, analysis and interpretation of data, and in the writing of the article.

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All authors have submitted the ICMJE Form for Disclosure of Potential Conflicts of Interest. Conflicts that the editors consider relevant to the content of the manuscript have been disclosed.

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