Review Article

Genetic Consequences of Antiviral Therapy on HIV-1

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Received 22 February 2015; Revised 26 May 2015; Accepted 27 May 2015

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A variety of enzyme inhibitors have been developed in combating HIV-1, however the fast evolutionary rate of this virus commonly leads to the emergence of resistance mutations that finally allows the mutant virus to survive. This review explores the main genetic consequences of HIV-1 molecular evolution during antiviral therapies, including the viral genetic diversity and molecular adaptation. The role of recombination in the generation of drug resistance is also analyzed. Besides the investigation and discussion of published works, an evolutionary analysis of protease-coding genes collected from patients before and after treatment with different protease inhibitors was included to validate previous studies. Finally, the review discusses the importance of considering genetic consequences of antiviral therapies in models of HIV-1 evolution that could improve current genotypic resistance testing and treatments design.

1. Introduction

According to UNAIDS, the Joint United Nations Programme on HIV/AIDS World Health Organization, a total of 35.3 [32.2–38.8] million people worldwide were living with HIV-1 in 2012, indicating a ∼15% increase of infected people from 2001 [1]. A total of 2.3 [1.9–2.7] million were newly infected during 2012, showing a 33% decline of new infections from 2001 with 3.4 [3.1–3.7] million. Indeed, the number of AIDS deaths declined from 2.3 [2.1–2.6] million in 2005 to 1.6 [1.4–1.9] million in 2012 [1]. An important cause for such a death decline is the antiretroviral therapy, usually referred to as highly active antiretroviral therapy (HAART). In 2012, a total of 9.7 million people from low/middle-income countries received HAART and the UNAIDS expects to reach 15 million people receiving HAART by 2015 [1]. Nevertheless, in 2013 only 34% of people infected with HIV in low/middle-income countries (28.6 million) could receive therapy [1]. Therefore, there are still important regional differences that should be solved [2, 3]. On the other hand, the development of an effective HIV vaccine is still under progress with a number of failures [4] because of the high rate of evolution of HIV-1 [5, 6]. As a consequence, up to date the only treatment for HIV-1 is the antiretroviral drug therapy.

HAART have largely delayed the onset of AIDS-related illness and death [1] although they cannot eradicate the virus mainly due to latent viral reservoirs [7]. In addition, drug resistance mutations can reduce the activity of the therapy [8, 9]. Drug resistance mutations probably emerge because HIV evolves rapidly, with high mutation and recombination rates and under rapid population dynamics [10]. Of course, then natural and drug-induced selection can eliminate most of viral variants [11]. The surviving variants (8–20%) present drug resistance mutations, which allows recovering fitness and replication capacity [8, 12]. Interestingly, different inhibitors can generate different selective pressures that induce the fixation of different resistance mutations in the viral population but also different resistance mutations may affect different inhibitors in a different fashion. This suggests the simultaneous use of more than one inhibitor that could cover a wider range of mutations [13], although this strategy may fall into similar resistance (cross-resistance) and lack of synergy [14, 15].

A potential strategy to deal with the problem of resistance mutations could be the consideration of the molecular evolution of the virus [16, 17] into the inhibitor design. For example, inhibitors that account for molecular evolutionary processes of the virus could eliminate viral variants that could
be predicted beforehand. Actually, this promising strategy is commonly applied to HIV-1 vaccines design through the use of centralized (consensus, center-of-tree or ancestral) genes that can induce immune responses (reviewed in [17]). Such centralized sequences could consider the immunogenetic particularities of the diverse circulating variants in the target population [18, 19]. However, although these centralized vaccines generated promising antibody responses, they were only partially effective in covering the large HIV-1 genetic diversity. Perhaps this could be derived from the application of not enough realistic models of HIV-1 evolution as suggested in [17]; see also [20]. Knowledge on HIV-1 molecular evolution can also be used to develop realistic models of evolution [21, 22] that can be applied for additional purposes such as the prediction of resistance mutations [23] or the evolutionary reply of the viral population, genotypic resistance testing [24, 25].

This study explores the genetic consequences of antiviral therapy on HIV-1. First, it analyzes the influences of antiviral therapies on the viral genetic diversity, including the particular roles of the substitution and recombination processes in the generation of drug resistance. Then, the molecular signatures of selective pressures derived from antiviral therapies are evaluated. A brief evolutionary analysis of the influence of different protease (PR) inhibitors (PIs) on the PR-coding region was performed to evaluate previous works and to provide an illustrative example. The application of the genetic consequences derived from antiviral therapies in the development of new empirical substitution models that could be used for purposes such as genotypic resistance testing and treatments design is also discussed.

2. Genetic Diversity Generated during HIV-1 Antiviral Therapy

Interestingly, the effects of HIV-1 antiviral drugs on the viral genetic diversity depend on the evolutionary level under study. It differs from overall diversity of circulating strains in the viral population to local nucleotide diversity of particular viral genes.

The antiviral therapy can reduce the global viral genetic diversity in the population due to the selection of viral strains [26–28]. This phenomenon can be interpreted as a classical population range contraction and habitat fragmentation that commonly tend to decrease genetic diversity [29, 30]. Actually, a variety of population genetics analysis of HIV-1 showed the existence of severe population bottlenecks (loss of viral load) and loss of virus fitness during drug regimens [28, 31, 32].

In contrast, the survival strains may present drug resistance mutations [33] that often increase genetic diversity of the protein-coding genes of the target proteins [12, 34, 35]. Wu et al. [36] found that patients treated with several PIs presented 3 times more protease mutations than untreated patients. These findings are also observed in the computational analysis presented in the last section of this paper where most of PIs promoted higher levels of nucleotide diversity in the PR-coding gene. Interestingly, pairs and clusters of correlated resistance mutations (coevolution) were significantly more abundant in treated patients [36]. Consequently, the increased diversity does not follow a random process, instead the new mutations present residue-residue interactions from direct association with viral protein inhibitors [23]. Increased genetic diversity can also be observed under treatment with other antiviral drugs such as reverse transcriptase (RT) inhibitors [34, 37–39] and integrase (IN) inhibitors [40–43], although the increased diversity under the latter drug class is mainly based on secondary resistance mutations [40–43]. Indeed, combinations of different drug classes (acting on different HIV-1 proteins) can generate synergistic inhibition [44] but the overall presence of synergy on genetic diversity remains to be explored, although some mutations in the Env region have already been associated with resistance to entry inhibitors that affect other viral genes [9]. Overall, at this level, the molecular mechanisms by which the virus can evade treatments seem directly related with the virus’s ability to generate genetic diversity in a particular environment. Thus, this increased genetic diversity could be driven by strong selective pressures (discussed later).

3. The Role of Viral Recombination during HIV-1 Antiviral Therapy

Recombination constitutes a fundamental evolutionary force in HIV generating new viral strains, increasing viral diversity, and facilitating adaptation [45–47]. Indeed, ignored recombination can bias the inference of a variety of evolutionary processes and parameters (i.e., it can increase the number of false positively selected sites [48, 49] or generate incorrect phylogenetic tree and ancestral sequence reconstructions [50, 51]). Therefore recombination should be taken into account for analyzing and understanding HIV-1 evolution.

The role of recombination on the emergence of drug resistance mutations is not yet clear and it can be difficult to assess because other processes may also influence its evolutionary consequences (i.e., cellular superinfection [52–54], random genetic drift, and viral population size [55, 56] or fitness selection of the newly generated viral forms [57, 58]) and because the detection of recombination can be problematic under low levels of nucleotide diversity [59]. Contradictory effects of recombination during HIV-1 antiviral therapy can be found in the literature.

As one would expect beforehand, several studies showed that recombination is crucial to generate drug resistance. A computer simulations study suggested that recombination might favor the generation of drug resistance [60]. In addition, HIV-1 strains derived from recombination events presented resistance mutations [61, 62].

On the contrary, Archer et al. [63] showed that despite the wide diversity of recombinant forms in HIV populations, only a minority of recombination events are of significance to the evolution of the virus. Counterintuitively, it has also been demonstrated that recombination can slow down the generation of multi-drug-resistant strains during therapy [52] and it may be suppressed by selection for resistance to PIs [64].

It seems that the initial genetic barrier caused by recombination (most of recombinant forms could present low
fitness) could reduce the fitness of the viral population during
the therapy but in case a recombinant form is selected, resistance mutations could be better able to persist in the viral population [54] and speed up adaptation (the Fisher-Muller effect) [65]. In any case, these opposite findings suggest that more sophisticated analyses should be performed to determine the influence of recombination on the emergence of drug resistance mutations, as suggested by Shi et al. [61].

4. Selective Pressures Induced by HIV-1 Antiviral Therapy

Antiviral therapy may cause important selective pressures on viral populations [12, 66]. In particular, severe fitness losses can be derived from antiviral treatments until the emergence of beneficial mutations that allow restoring the vital replication capacity [12]. Thus, resistance to viral inhibitors can drive the fixation of favorable variants [23, 67].

The overall response to antiviral drugs presented an excess of nonsynonymous substitutions [23, 68] (which was also found in the analysis presented in the following section). For example, Wu et al. [36] found that an antiviral therapy can induce diversifying selection in nearly one-half of PR sites. It is widely known that positively selected sites (PSSs) are often located in the protein surface, whereas conserved or negatively selected sites (NSSs) are commonly observed in the protein core in order to conserve the protein function [69]. However, the molecular adaptation induced by antiviral therapies does not present such a scenario. Poon et al. [23] found that the distribution of nonsynonymous substitutions along the gene is shaped by selection to PI resistance. Moreover, antiviral therapies promote complex drug-specific residue-residue interaction networks [23, 70, 71] that can drive the coevolution of primary and secondary resistance mutations [8, 23].

5. Genetic Impact of Diverse PIs on HIV-1 PR-Coding Genes: A Computational Study

The HIV-1 PR is one of the most used drug targets for combating HIV with a number of chemically diverse inhibitors that have already been tested [72, 73]. This section includes a computational analysis of nucleotide diversity and molecular adaptation of the PR-coding gene evolution under different PIs.

5.1. Sample Collection. Samples of coding DNA sequences that encode the HIV-1 PR (Pol region, subtype B) were collected from the Stanford HIV Drug Resistance Database [74, 75]. Subtype B was used because most (~99%) of datasets available in the database belong to this subtype and there is not enough data to analyze other subtypes. For each HIV-1 patient, a clonal sequence was collected under no-treatment and another one was collected after a particular treatment based on a single PI or a PIs combination. According to the detailed information provided by the database [74, 75], the patients did not receive other treatments. Therefore, to study each treatment (hereafter, evolutionary scenario) two datasets (pool of sequences before and after treatment) were obtained. In particular, for each evolutionary scenario, a dataset includes coding sequences collected before a given treatment and the other dataset includes coding sequences collected after such a treatment, and both datasets come from the same patients. As suggested by Kosakovsky Pond and Frost [76], scenarios with sample size lower than 10 were not considered to avoid lack of power in the evolutionary analysis (databases with higher sample size can generate accurate estimates of genetic diversity and nonsynonymous to synonymous substitution rates ratio (dN/dS) [76]; see also [3, 77]). A total of 13 evolutionary scenarios, all the currently available scenarios from the database, were analyzed. Namely, a “control” scenario (no-treatment in both datasets, scenario 1, 1011 patients) and scenarios with the following treatments: amprenavir (APV, scenario 2, 15 patients), atazanavir (ATV, scenario 3, 23 patients), indinavir (IDV, scenario 4, 77 patients), lopinavir (LPV, scenario 5, 34 patients), nelfinavir (NFV, scenario 6, 317 patients), ritonavir (RTV, scenario 7, 24 patients), saquinavir (SQV, scenario 8, 35 patients), and the PI combinations: IDV + RTV (scenario 9, 10 patients), RTV + SQV (scenario 10, 11 patients), and IDV + RTV + SQV (scenario 11, 11 patients). Two additional scenarios were also studied where patients treated with IDV are then treated with IDV + NFV (scenario 12, 13 patients) or IDV + RTV (scenario 13, 16 patients).

5.2. Analysis of Genetic Diversity and Recombination. Several genetic statistics were applied to study the influence of PIs on the genetic diversity of the PR-coding gene. (i) The overall sequences divergence was computed with MEGA 6.0 [78]. (ii) Nucleotide diversity (\(\pi\)) was estimated by using the pairwise nucleotide differences per site [79]. These metrics considered indels as missing data. (iii) The genetic distance between the two datasets of each evolutionary scenario was computed by the Kullback-Leibler (KL) divergence [80] and considering indels as missing data and as an additional state. This distance provides a comparative analysis of nucleotide diversity distributions across sites between two datasets [81].

Briefly, the results show that almost all PIs lead to higher levels of sequences divergence, pairwise nucleotide diversity, and nucleotide diversity distribution across sites. Except for LPV and NFV, all PIs increased the overall difference between sequences (Figure 1(a)). Similar results are derived from the estimates of nucleotide diversity although here only LPV presented low levels of nucleotide diversity variation (see Figure S1 in Supplementary Material available online at http://dx.doi.org/10.1155/2015/395826). The highest levels of diversity were generated from treatment with APV, IDV, and, especially, PIs combinations. However, the increase of diversity could be caused by the emergence of resistance mutations but also by mutations derived from the natural evolution of the gene. Therefore, it is interesting to evaluate the correlation between the variation of diversity and the corresponding time period between samples. Figure 2(a) suggests that there is no correlation between these parameters, which is supported by a low correlation coefficient \(r = 0.056\). For example, the control dataset (no-treatment) does not present increase of diversity despite its long time period (11 months), whereas treatment with APV generated one of the highest
levels of diversity in only 4 months (Figure 2). In addition, correlation coefficients within scenarios (among patients from a particular scenario) were also very low, most of them under 0.1 (Figure 2(a)). A normalization dividing the genetic diversity gradient by the time period between samples also indicated the increase of genetic diversity with most of PIs (Figure 2). However, the normalization must be carefully interpreted because a longer time period does not necessarily lead to more diversity [82], which is actually indicated by the described lack of correlation.

The analysis of nucleotide diversity distribution across sites between the two alignments showed similar findings for most of PIs (Figure 1(b)). Notice that this nucleotide distribution can be more influenced by several mutations at specific positions [80] and therefore this statistic might be more sensible to detect resistance mutations. The results show an influence of all PIs on the nucleotide diversity distribution (Figure 1(b)), although this influence varies among inhibitors. Again, the long KL distance derived from drug combination therapies is remarkable.

Absence of recombination breakpoints was found with the single breakpoint position (SBP) method [83], implemented in the Hyphy package [84], and with the recombination detection methods implemented in the RDP framework [85].

5.3. Signatures of Molecular Adaptation. The best-fit model of DNA substitution was selected with jModelTest [86] under the Bayesian information criterion (BIC), as suggested by [87]. Then, maximum likelihood (ML) phylogenetic trees were inferred with PhyML [88] under the corresponding substitution model. These trees were used to perform the molecular adaptation inferences. Estimates of \( dN/dS \) at both global (sequence) and local (codon) levels were performed

**Figure 1:** Overall sequence diversity variation and Kullback-Leibler divergence. (a) Variation of overall sequence difference between the two datasets of each evolutionary scenario \( (d_{after \ treatment} - d_{before \ treatment}) \). Indels are considered as missing data. Error bars indicate standard error. Reference values are shown in Table S1 (Supplementary Material). (b): Kullback-Leibler distance, nucleotide diversity distribution, between the two datasets of each evolutionary scenario. Dark grey bars consider indels as a new state whereas clear grey bars consider indels as missing data. Error bars indicate standard error across sites. “−” indicates naïve-treatment patients.

**Figure 2:** Sequence difference variation as a function of time interval between samples. (a) Variation of the overall sequence difference between the two datasets of each evolutionary scenario \( (d_{after \ treatment} - d_{before \ treatment}) \) is represented in the “y-axis” (mean and standard error). The time period between both samples \( (t_{after \ treatment} - t_{before \ treatment}) \) is represented in the “x-axis” (mean and standard error from all patients of the scenario). The correlation coefficient between both parameters among all the scenarios is \( r = 0.056 \), suggesting absence of correlation. Correlation coefficients within each evolutionary scenario (among patients) between these parameters are also shown in the plot and ranges from 0.0003 to 0.507, although most of them are under 0.1. (b) Genetic diversity gradient divided by the corresponding time period. Error bars indicate standard error. “−” indicates naïve-treatment patients. Reference values are shown in Table S1 (Supplementary Material).
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The evolutionary analysis included in this work supported such considerations. It also showed that different PIs can promote different influences on the genetic diversity of the viral PR-coding gene. For example, IDV and APV induced the highest levels of diversity (among treatments with a single PI) and PIs combination induced very high levels of genetic diversity, especially when 3 PIs are applied jointly. As noted, this increase of diversity can be related with the emergence of resistance mutations [33]. The reason why some inhibitors induce more diversity than others requires complex structural analysis of enzyme-inhibitor interactions, which is an important topic of research [89, 90]. On the other hand, absence of recombination breakpoints was found in the analyzed PR-coding genes. This could be caused by limitations to detect recombination under low genetic diversity levels [59] or just because recombination was not required to generate drug resistance. Indeed, recombination could occur in other genomic regions [63] (i.e., recombinants with breakpoints in Gag and Pol may present selection against [63, 91]). Concerning molecular adaptation, the results showed that the wild evolution of the virus presents an overall decrease of the global \( dN/dS \) without PSSs and where most of sites evolved under significant purifying selection. This is probably caused by the sharp purifying selection induced by the host’s immune system. On the other hand, PIs often promoted an overall increase of the global \( dN/dS \) (see [23, 68] and Figure 3), which is most of times accompanied by the emergence of significant PSSs along the gene (see [23] and Figure S2). As expected, the largest \( dN/dS \) increase occurs under treatments with PIs combination (see [23] and Figure 3). These signatures of molecular adaptation are related with the amount of genetic diversity induced by the PIs and indicate the primary importance of adaptation in the evolutionary process of the PR-coding gene under PIs. On the other hand, a large number of codons evolved under negative (purifying) selection, which indicates the presence of strong selective pressures, as noted probably caused by the host’s immune system and the therapy.

Understanding molecular evolution of the virus can help us develop more realistic models of HIV evolution [23, 69, 71], correlate the disease progression with the evolution of the viral population [28, 92], and predict resistance (i.e., by genotypic-resistance testing [24, 25, 93, 94]) and common ancestry [18], or vaccine design [17–19]. Nevertheless, HIV-1 evolution is complex and other phenomena should also be taken into account as much as possible in the models, for example, different host’s immune responses, clinical stage, HIV-1 compartmentalization [95], or infection with multiple viral variants, although the latter presents an overall low incidence [53, 96].

Since antiviral therapies affect genetic diversity of the virus by strong selective pressures, models of HIV-1 evolution should accommodate such effects in order to mimic these scenarios for purposes such as robust genotypic resistance testing and treatments design. Importantly, models of HIV-1 evolution should be as realistic as possible in order to provide accurate predictions. A possibility could be the consideration of a fitness landscape (e.g., [97]) to develop parametric models. However, the design and computation of a realistic fitness function are too convoluted due to complex processes with the Fixed Effects Likelihood (FEL) method [76] implemented in the Hyphy package. Notice that this ML-based method provides very accurate estimates [76] and it is commonly used in population genetics and virus evolution (e.g., [3]).

Figure 3 shows the variation of global \( dN/dS \) estimates from datasets collected before and after a treatment. All the PIs promoted increased estimates of \( dN/dS \), especially when the treatment is based on PIs combination. By contrast, in absence of treatment the estimated \( dN/dS \) declined with time. At the local level, almost all the PIs promoted an increase of significant (\( p \) value < 0.05) PSSs along the PR-coding sequence (Figure S2 and Table S4, Supplementary Material). In general, a large number of NSSs were detected in all datasets (Table S4) without showing a clear relationship with the presence or absence of treatment. All these results are discussed in the next section.

6. Concluding Remarks

The fast population range contractions and fragmentation produced during the therapy can reduce the overall diversity of viral strains [29] and, by contrast, the emergence of resistance mutations caused from the rapid evolution of HIV allows preserving or increasing the levels of nucleotide diversity of viral protein-coding genes of the drug target. Indeed, resistance mutations can be rare, but also recurrent enough until they reach resistance, and can generate positive selection driving the fixation of favorable viral strains [23, 67]. At this level two opposite selective pressures seem to act. While most of sites evolve under strong purifying selection, probably caused by the host’s immune system and the therapy, other sites evolve under diversifying selection, probably caused by the viral molecular adaptation to the new environment established by the therapy, and can present complex residue-residue interaction networks suggesting dependent evolution among sites [23, 70].
that affect viral genetic diversity such as antiviral therapies (as noted in this paper). An easier, but less robust, alternative can be the development of scenario-specific empirical models. As shown in this paper, different therapies must be modeled with different models of evolution since different therapies can promote different genetic consequences in the virus. Much more research is needed (i.e., the consideration of associations between observed genotypes and phenotypic resistance in models of HIV-1 evolution) but my impression is that HIV-1 therapies will benefit from more consideration of evolutionary information.

Conflict of Interests

The author declares that there is no conflict of interests regarding the publication of this paper.

Acknowledgments

The author wants to thank three anonymous reviewers for insightful comments. This work was supported by the Spanish Government through the “Juan de la Cierva” Fellowship ICI-2011-10452 and by the Portuguese Government through the FCT Starting Grant IF/00955/2014.

References

[1] UNAIDS, “2013 report on the global AIDS epidemic,” in Joint United Nations Programme on HIV/AIDS, World Health Organization, 2013.
[2] B. S. Peters and K. Conway, “Therapy for HIV: past, present, and future,” Advances in Dental Research, vol. 23, no. 1, pp. 23–27, 2011.
[3] M. Pérez-Losada, D. Posada, M. Arenas et al., “Ethnic differences in the adaptation rate of HIV gp120 from a vaccine trial,” Retrovirology, vol. 6, article 67, 2009.
[4] A. McMichael, L. J. Picker, J. P. Moore, and D. R. Burton, “Another HIV vaccine failure: where to next?” Nature medicine, vol. 19, no. 12, pp. 1576–1577, 2013.
[5] K. S. Slobod, M. Bonsignori, S. A. Brown, X. Zhan, J. Stambas, and J. L. Hurwitz, “HIV vaccines: brief review and discussion of future directions,” Expert Review of Vaccines, vol. 4, no. 3, pp. 305–313, 2005.
[6] B. T. Korber, N. L. Letvin, and B. F. Haynes, “T-cell vaccine strategies for human immunodeficiency virus, the virus with a thousand faces,” Journal of Virology, vol. 83, no. 17, pp. 8300–8314, 2009.
[7] J. D. Siliciano and R. F. Siliciano, “The latent reservoir for HIV-1 in resting CD4+ T cells: a barrier to cure,” Current Opinion in HIV and AIDS, vol. 1, no. 2, pp. 121–128, 2006.
[8] L. Menéndez-Arias, “Molecular basis of human immunodeficiency virus type 1 drug resistance: overview and recent developments,” Antiviral Research, vol. 98, no. 1, pp. 93–120, 2013.
[9] A. M. Wensing, V. Calvez, H. E. Gunthard et al., “2014 Update of the drug resistance mutations in HIV-1,” Topics in Antiviral Medicine, vol. 22, no. 5, pp. 642–650, 2014.
[10] D. Shriner, A. G. Rodrigo, D. C. Nickle, and J. I. Mullins, “Pervasive genomic recombination of HIV-1 in vivo,” Genetics, vol. 167, no. 4, pp. 1573–1583, 2004.
[27] S. Harada, K. Yoshimura, A. Yamaguchi, S. Boonchawalit, K. Yusa, and S. Matsushita, “Impact of antiretroviral pressure on selection of primary human immunodeficiency virus type 1 envelope sequences in vitro,” Journal of General Virology, vol. 94, part 5, pp. 933–943, 2013.

[28] R. M. Troyer, K. R. Collins, A. Abraha et al., “Changes in human immunodeficiency virus type 1 fitness and genetic diversity during disease progression,” Journal of Virology, vol. 79, no. 14, pp. 9006–9018, 2005.

[29] M. Arenas, N. Ray, M. Cururat, and L. Excoffier, “Consequences of range contractions and range shifts on molecular diversity,” Molecular Biology and Evolution, vol. 29, no. 1, pp. 207–218, 2012.

[30] S. Mona, N. Ray, M. Arenas, and L. Excoffier, “Genetic consequences of habitat fragmentation during a range expansion,” Heredity, vol. 112, no. 3, pp. 291–299, 2014.

[31] K. M. Kitrinos, J. A. E. Nelson, W. Resch, and R. Swanstrom, “Effect of a protease inhibitor-induced genetic bottleneck on human immunodeficiency virus type 1 env gene populations,” Journal of Virology, vol. 79, no. 16, pp. 10627–10637, 2005.

[32] A. Ibáñez, B. Clotet, and M.-A. Martínez, “Human immunodeficiency virus type 1 population bottleneck during inactivin therapy causes a genetic drift in the env quasispecies,” Journal of General Virology, vol. 81, no. 1, pp. 85–95, 2000.

[33] S. A. Rabi, G. M. Laird, C. M. Durand et al., “Multi-step inhibition explains HIV-1 protease inhibitor pharmacodynamics and resistance,” Journal of Clinical Investigation, vol. 123, no. 9, pp. 3848–3860, 2013.

[34] P. Zhong, Q. Pan, Z. Ning et al., “Genetic diversity and drug resistance of human immunodeficiency virus type 1 (HIV-1) strains circulating in Shanghai,” AIDS Research and Human Retroviruses, vol. 23, no. 7, pp. 847–856, 2007.

[35] I. Chen, L. Khaki, J. C. Lindsey et al., “Association of pol diversity with antiretroviral treatment outcomes among HIV-infected African children,” PLoS ONE, vol. 8, no. 11, Article ID e81213, 2013.

[36] T. D. Wu, C. A. Schiffer, M. J. Gonzales et al., “Mutation patterns and structural correlates in human immunodeficiency virus type 1 protease following different protease inhibitor treatments,” Journal of Virology, vol. 77, no. 8, pp. 4836–4847, 2003.

[37] R. W. Shafer, R. Kantor, and M. J. Gonzales, “The genetic basis of HIV-1 resistance to reverse transcriptase and protease inhibitors,” AIDS Reviews, vol. 2, no. 4, pp. 211–228, 2000.

[38] D. J. de Sa-Filho, M. D. S. Soares, V. Candido et al., “HIV type 1 pol gene diversity and antiretroviral drug resistance mutations in Santos, Brazil,” AIDS Research and Human Retroviruses, vol. 24, no. 3, pp. 347–353, 2008.

[39] R. Kantor and D. Katzenstein, “Polymorphism in HIV-1 non-subtype b protease and reverse transcriptase and its potential impact on drug susceptibility and drug resistance evolution,” AIDS Reviews, vol. 5, no. 1, pp. 25–35, 2003.

[40] L. B. Arruda, L. A. M. Fonseca, A. J. S. Duarte, and J. Casseb, “Genetic diversity on the integrase region of the pol gene among HIV type 1-infected patients naïve for integrase inhibitors in São Paulo City, Brazil,” AIDS Research and Human Retroviruses, vol. 26, no. 1, pp. 105–107, 2010.

[41] G. S. Gottlieb, R. A. Smith, N. M. D. Badiane et al., “HIV-2 integrase variation in integrase inhibitor-naïve adults in Senegal, West Africa,” PLoS ONE, vol. 6, no. 7, Article ID e22204, 2011.

[42] C. Garrido, A. M. Geretti, N. Zahonero et al., “Integrate variability and susceptibility to HIV integrase inhibitors: impact of subtypes, antiretroviral experience and duration of HIV infection,” The Journal of Antimicrobial Chemotherapy, vol. 65, no. 2, pp. 320–326, 2010.

[43] A. Piralla, S. Paolucci, R. Gulminetti, G. Comolli, and F. Baldanti, "HIV integrase variability and genetic barrier in antiretroviral naive and experienced patients," Virology Journal, vol. 8, article 149, 2011.

[44] R. Kulkarni, R. Hluhanich, D. M. McColl, M. D. Miller, and K. L. White, "The combined anti-HIV-1 activities of emtricitabine and tenofovir plus the integrase inhibitor elvitegravir or raltegravir show high levels of synergy in vitro," Antimicrobial Agents and Chemotherapy, vol. 58, no. 10, pp. 6145–6150, 2014.

[45] R. Nájera, E. Delgado, L. Pérez-Alvarez, and M. M. Thomson, "Genetic recombination and its role in the development of the HIV-1 pandemic," AIDS, vol. 16, supplement 4, pp. S3–S16, 2002.

[46] D. Moradigaravand, R. Kouyos, T. Hinkley et al., "Recombination accelerates adaptation on a large-scale empirical fitness landscape in HIV-1," PLoS Genetics, vol. 10, no. 6, Article ID e1004439, 2014.

[47] M. Pérez-Losada, M. Arenas, J. C. Galán, F. Palero, and F. González-Candelas, "Recombination in viruses: mechanisms, methods of study, and evolutionary consequences," Infection, Genetics and Evolution, vol. 30, pp. 296–307, 2015.

[48] M. Anisimova, R. Nielsen, and Z. Yang, "Effect of recombination on the accuracy of the likelihood method for detecting positive selection at amino acid sites," Genetics, vol. 164, no. 3, pp. 1229–1236, 2003.

[49] M. Arenas and D. Posada, "The influence of recombination on the estimation of selection from coding sequence alignments," in Natural Selection: Methods and Applications, M. A. Fares, Ed., pp. 112–125, CRC Press/Taylor & Francis, Boca Raton, Fla, USA, 2014.

[50] M. H. Schierup and J. Hein, "Consequences of recombination on traditional phylogenetic analysis," Genetics, vol. 156, no. 2, pp. 879–891, 2000.

[51] M. Arenas and D. Posada, "The effect of recombination on the reconstruction of ancestral sequences," Genetics, vol. 184, no. 4, pp. 1133–1139, 2010.

[52] M. T. Bretschger, C. L. Althaus, V. Müller, and S. Bonhoeffer, "Recombination in HIV and the evolution of drug resistance: for better or for worse?" BioEssays, vol. 26, no. 2, pp. 180–188, 2004.

[53] M. Pernas, C. Casado, R. Fuentes, M. J. Pérez-Elias, and C. López-Galindo, "A dual superinfection and recombination within HIV-1 subtype B 12 years after primoinfection," Journal of Acquired Immune Deficiency Syndromes, vol. 42, no. 1, pp. 12–18, 2006.

[54] C. Fraser, "HIV recombination: what is the impact on antiretroviral therapy?" Journal of the Royal Society Interface, vol. 2, no. 5, pp. 489–503, 2005.

[55] N. N. V. Vijay, R. Ajmani, A. S. Perelson, and N. M. Dixit, "Recombination increases human immunodeficiency virus fitness, but not necessarily diversity," Journal of General Virology, vol. 89, no. 6, pp. 1467–1477, 2008.

[56] D. Shriman, R. Shankarappa, M. A. Jensen et al., "Influence of random genetic drift on human immunodeficiency virus type 1 env evolution during chronic infection," Genetics, vol. 166, no. 3, pp. 1155–1164, 2004.

[57] L. J. Costa, P. Munerato, R. S. Díaz, and A. Tanuri, "Generation of intersubtype human immunodeficiency virus type 1 recombinants in env gene in vitro: influences in the biological behavior
and in the establishment of productive infections,” Virology, vol. 268, no. 2, pp. 440–451, 2000.

[58] M. E. Quiñones-Mateu, Y. Gao, S. C. Ball, A. J. Marozsan, A. Abrahá, and E. J. Arts, “In vitro intersubtype recombinants of human immunodeficiency virus type 1: comparison to recent and circulating in vivo recombinant forms,” Journal of Virology, vol. 76, no. 19, pp. 9600–9613, 2002.

[59] D. Posada and K. A. Crandall, “Evaluation of methods for detecting recombination from DNA sequences: computer simulations,” Proceedings of the National Academy of Sciences of the United States of America, vol. 98, no. 24, pp. 13757–13762, 2001.

[60] A. Carvajal-Rodriguez, K. A. Crandall, and D. Posada, “Recombination favors the evolution of drug resistance in HIV-1 during antiretroviral therapy,” Infection, Genetics and Evolution, vol. 7, no. 4, pp. 476–483, 2007.

[61] B. Shi, C. Kitchen, B. Weiser et al., “Evolution and recombination of the HIV-1 drug resistance and tropism during antiretroviral therapy,” Virology, vol. 404, no. 1, pp. 5–20, 2010.

[62] C. L. Althaus and S. Bonhoeffer, “Stochastic interplay between mutation and recombination during the acquisition of drug resistance mutations in human immunodeficiency virus type 1,” Journal of Virology, vol. 79, no. 21, pp. 13572–13578, 2005.

[63] J. Archer, J. W. Pinney, J. Fan et al., “Identifying the important HIV-1 recombination breakpoints,” PLoS Computational Biology, vol. 4, no. 9, Article ID e1000718, 7 pages, 2008.

[64] M. Nijhuis, C. A. B. Boucher, P. Schiper, T. Leitner, R. Schuurman, and J. Albert, “Stochastic processes strongly influence HIV-1 evolution during suboptimal protease-inhibitor therapy,” Proceedings of the National Academy of Sciences of the United States of America, vol. 95, no. 24, pp. 14441–14446, 1998.

[65] J. Kim and H. A. Orr, “Adaptation in sexuals vs. asexuals: clonal interference and the Fisher-Muller model,” Genetics, vol. 171, no. 3, pp. 1377–1386, 2005.

[66] M. Foll, Y.-P. Poh, N. Renzette et al., “Influenza virus drug resistance: a time-sampled population genetics perspective,” PLoS Genetics, vol. 10, no. 2, Article ID e1004185, 2014.

[67] A. E. Y. Poon, S. L. Kosakovsky Pond, P. Bennett, D. D. Richman, A. J. Leigh Brown, and S. D. W. Frost, “Adaptation to human populations is revealed by within-host polymorphisms in HIV-1 and hepatitis C virus,” PLoS Pathogens, vol. 3, no. 3, article e45, 2007.

[68] C. Pan, J. Kim, L. Chen, Q. Wang, and C. Lee, “The HIV positive selection mutation database,” Nucleic Acids Research, vol. 35, supplement 1, pp. D371–D375, 2007.

[69] J. Woo, D. L. Robertson, and S. C. Lovell, “Constraints on HIV-1 diversity from protein structure,” Journal of Virology, vol. 84, no. 24, pp. 12995–13003, 2010.

[70] M. John, C. B. Moore, I. R. James, and S. A. Mallal, “Interactive selective pressures of HLA-restricted immune responses and antiretroviral drugs on HIV-1,” Antiviral Therapy, vol. 10, no. 4, pp. 551–555, 2005.

[71] N. Beerenwinkel, B. Schmidt, H. Walter et al., “Diversity and complexity of HIV-1 drug resistance: a bioinformatics approach to predicting phenotype from genotype,” Proceedings of the National Academy of Sciences of the United States of America, vol. 99, no. 12, pp. 8271–8276, 2002.

[72] M. Arenas, M. C. Villaverde, and F. Sussman, “Prediction and analysis of binding affinities for chemically diverse HIV-1 PR inhibitors by the modified SAFE-p approach,” Journal of Computational Chemistry, vol. 30, no. 8, pp. 1229–1240, 2009.

[73] A. M. J. Wensing, N. M. van Maarseveen, and M. Nijhuis, “Fifteen years of HIV Protease Inhibitors: raising the barrier to resistance,” Antiviral Research, vol. 85, no. 1, pp. 59–74, 2010.

[74] S. Y. Rhee, M. J. Gonzales, R. Kantor, B. J. Betts, J. Ravela, and R. W. Shafer, “Human immunodeficiency virus reverse transcriptase and protease sequence database,” Nucleic Acids Research, vol. 31, no. 1, pp. 298–303, 2003.

[75] R. W. Shafer, “Rationale and uses of a public HIV drug-resistance database,” Journal of Infectious Diseases, vol. 194, supplement 1, pp. S51–S58, 2006.

[76] S. L. Kosakovskv Pond and S. D. W. Frost, “Not so different after all: a comparison of methods for detecting amino acid sites under selection,” Molecular Biology and Evolution, vol. 22, no. 5, pp. 1208–1222, 2005.

[77] R. Batzorsky, M. F. Kearney, S. E. Palmer, F. Maldarelli, I. M. Rouzine, and J. M. Coffin, “Estimate of effective recombination rate and average selection coefficient for HIV in chronic infection,” Proceedings of the National Academy of Sciences of the United States of America, vol. 108, no. 14, pp. 5661–5666, 2011.

[78] K. Tamura, G. Stecher, D. Peterson, A. Filipski, and S. Kumar, “MEGA6: Molecular evolutionary genetics analysis version 6.0,” Molecular Biology and Evolution, vol. 30, no. 12, pp. 2725–2729, 2013.

[79] M. Nei, Molecular Evolutionary Genetics, Columbia University Press, New York, NY, USA, 1987.

[80] S. Kullback, “Letter to the editor: the Kullback-Leibler distance,” The American Statistician, vol. 41, no. 4, pp. 340–341, 1987.

[81] M. Arenas, H. G. Dos Santos, D. Posada, and U. Bastolla, “Protein evolution along phylogenetic histories under structurally constrained substitution models,” Bioinformatics, vol. 29, no. 23, pp. 3020–3028, 2013.

[82] G. Bello, C. Casado, S. García et al., “Lack of temporal structure in the short term HIV-1 evolution within asymptomatic naive patients,” Virology, vol. 362, no. 2, pp. 294–303, 2007.

[83] S. L. K. Pond, D. Posada, M. B. Gravenor, C. H. Woelk, and S. D. W. Frost, “Automated phylogenetic detection of recombination using a genetic algorithm,” Molecular Biology and Evolution, vol. 23, no. 10, pp. 1891–1901, 2006.

[84] S. L. Kosakovskv Pond, S. D. W. Frost, and S. V. Muse, “HyPhy: hypothesis testing using phylogenies,” Bioinformatics, vol. 21, no. 5, pp. 676–679, 2005.

[85] D. P. Martin, P. Leme, M. Lott, V. Moulton, D. Posada, and P. Lefeuvre, “RDP3: a flexible and fast computer program for analyzing recombination,” Bioinformatics, vol. 26, no. 19, Article ID btq467, pp. 2462–2463, 2010.

[86] D. Posada, “ModelTest: phylogenetic model averaging,” Molecular Biology and Evolution, vol. 25, no. 7, pp. 1253–1256, 2008.

[87] A. Luo, H. Qiao, Y. Zhang et al., “Performance of criteria for selecting evolutionary models in phylogenetics: a comprehensive study based on simulated datasets,” BMC Evolutionary Biology, vol. 10, no. 1, article 242, 2010.

[88] S. Guindon, J.-F. Dufayard, V. Lefort, M. Anisimova, W. Hordijk, and O. Gascuel, “New algorithms and methods to estimate maximum-likelihood phylogenies: assessing the performance of PhyML 3.0,” Systematic Biology, vol. 59, no. 3, pp. 307–321, 2010.

[89] T. Hinkley, J. Martins, C. Chappey et al., “A systems analysis of mutational effects in HIV-1 protease and reverse transcriptase,” Nature Genetics, vol. 43, no. 5, pp. 487–490, 2011.
[90] G. J. Henderson, S.-K. Lee, D. M. Irlbeck et al., “Interplay between single resistance-associated mutations in the HIV-1 protease and viral infectivity, protease activity, and inhibitor sensitivity,” *Antimicrobial Agents and Chemotherapy*, vol. 56, no. 2, pp. 623–633, 2012.

[91] D. P. Martin, E. van der Walt, D. Posada, and E. P. Rybicki, “The evolutionary value of recombination is constrained by genome modularity,” *PLoS genetics*, vol. 1, no. 4, p. e51, 2005.

[92] M. A. Nowak, R. M. Anderson, M. C. Boerlijst, S. Bonhoeffer, R. M. May, and A. J. McMichael, “HIV-1 evolution and disease progression,” *Science*, vol. 274, no. 5289, pp. 1008–1011, 1996.

[93] B. Zöllner, H.-H. Feucht, L. Weitner et al., “Application of HIV-1 genotypic-resistance testing prevents the evolution of further resistance mutations in heavily pretreated patients,” *Journal of Clinical Virology*, vol. 21, no. 1, pp. 37–45, 2001.

[94] M. L. Branham, E. A. Ross, and T. Govender, “Predictive models for maximum recommended therapeutic dose of antiretroviral drugs,” *Computational and Mathematical Methods in Medicine*, vol. 2012, Article ID 469769, 9 pages, 2012.

[95] G. Tirado, G. Jove, R. Kumar et al., “Compartmentalization of drug resistance-associated mutations in a treatment-naive HIV-infected female,” *AIDS Research and Human Retroviruses*, vol. 20, no. 6, pp. 684–686, 2004.

[96] D. M. Smith, J. K. Wong, G. K. Hightower et al., “Incidence of HIV superinfection following primary infection,” *Journal of the American Medical Association*, vol. 292, no. 10, pp. 1177–1178, 2004.

[97] R. Lorenzo-Redondo, S. Delgado, F. Morán, and C. Lopez-Galindez, “Realistic three dimensional fitness landscapes generated by self organizing maps for the analysis of experimental HIV-1 evolution,” *PLoS ONE*, vol. 9, no. 2, Article ID e88579, 2014.
