Prediction of High Level of Multiple Drug Resistance Mutations in HIV-1 Subtype C Reverse Transcriptase Gene among First Line Antiretroviral-Experienced Virological Failure Patients from North India Using Genotypic and Docking Analysis

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Abstract

Background: Virologic failures and development of drug resistance can result in reduced treatment options in HIV infection.

Methods: RT sequence of HIV-1 subtype C isolates from 122 Antiretroviral Therapy (ART) naive and 13 virological and first line regimen failures from North India were analyzed. Mutations were defined according to Stanford Drug Resistance data base. A three dimensional HIV-1 subtype C specific computational model of RT was created from consensus sequences from naïve patients to analyze mutations in therapy failures. CD4 count and viral load were measured to analyze the disease status and subtyping was done using Genotyping NCBI HIV subtyping tool.

Results: All thirteen isolates from first-line ART-failure patients had mutations effecting susceptibility to RTI drugs when analyzed using Stanford DR HIV-1 database. The most common NRTI resistance mutations were in positions 118, 184, 210 and 215 indicating the possibility of high level resistance to Lamivudine (3TC) and Emtricitabine (FTC) in 92.3% of isolates. Common NNRTI resistance mutations were identified at position 98, 101, 103, 181 and 190 indicating a high level resistance to Nevirapine (NVP) in 100% therapy failures, while affecting the susceptibility to EFV in 76.92%. Energy scores were calculated after docking of various NRTIs on our newly proposed model based on the three dimensional structure of local wild type reverse transcriptase (RT) of subtype C. The presence of V75M mutation in one of the isolates (SK-206) seem to be partially neutralizing the resistance effect of mutations 118I, 184V, 210W, and 215Y for Stavudine (D4T), Didanosine (DDI) and FTC while 75M decreases the susceptibility of Zidovudine (AZT) (ΔG=0.87), causing high level of resistance.

Conclusion: The data suggest that the proposed model was successful in predicting the resistance/susceptibility to various RTIs based on docking energy scores taking into consideration the cumulative effect of all the mutations together.

Keywords: HIV-1 subtype C; Therapy failures; Sequencing; Docking analysis; Molecular modeling

Introduction

Acquired Immunodeficiency Syndrome (AIDS), caused by the human immunodeficiency virus (HIV), is one of the leading causes of death with major medical and economic impact on the society. Due to high mutation rates associated with RNA replication and retro-transcription, there is spontaneous emergence of a pool of mutant viruses, some of which may be associated with drug resistance (DR) [1-3]. Keeping in mind the increasing demand of improving the drug efficacy, there is an urge to improve the methodology of identification of these mutations and know how they affect the susceptibility to a particular drug.

Subtype C of the HIV-1 is accountable for over 50% of the HIV-1 infections in Southeast Asia and African Countries while subtype B predominates in Western Europe and North America [4-7]. An increasing body of experimental evidence suggests that different HIV-1 subtypes exhibit disparate biological behaviors, and might respond differently to diagnostic, immunologic and therapeutic interventions [8,9]. Recent studies have identified subtype specific differences in HIV susceptibility to specific anti-retroviral drugs [10,11] and signature mutations selected by treatment [12-14]. An increased amount of resistance in subtype C HIV has been documented as compared to other subtypes in response to single-dose NVP therapy [15,16]. This indicates that subtype of HIV might have influence on the level of resistance development against a particular anti-retroviral drug under that drug pressure and deserves further attention to prove this fact. For genotypic analysis of isolates from infected individuals the sequences need to be analyzed using one of the few drug-resistance databases available which include Stanford DR database, the Los Alamos HIV sequence database, the Rega algorithm for HIV subtype analysis, International AIDS society drug-resistance information to name a few. All these databases are mainly based on subtype-B sequences. In order to have a better
interpretation of genotype variants in a subtype-C related infections with respect to their behavior in response to existing drug regimens, we have a proposed a computational model of subtype C specific wild type RT constructed after introducing the genetic variations. Mutated HIV-1 subtype C HIV-1 RT and inhibitor complexes were created after docking various nucleotide reverse transcriptase inhibitors (NRTIs) and binding energy was calculated for each. The data suggest that best drug combinations for a given set of mutations in subtype C infected individuals can be designed using the information generated from our proposed model, which was capable of calculating the combinatorial effect of all the mutations in that particular isolate, rather than effects in isolation for each mutation.

Material and Methods

Patient selection and sample collection

For the analysis of Drug Resistance (DR) mutations and genetic diversity in the RT gene of HIV-1 subtype C strains in North India, 122 HIV-1 positive drug naïve individuals and 13 patients who had received first-line therapy (combination of 2 NRTIs and one NNRTI) for 4-5 years and shown clinical as well as virological signs of treatment failure (HIV-1 RNA >1000 copies/ml) were recruited. The study was approved by the Institutional Ethics Committee. After informed consent, 3 ml of venous blood was drawn in a K3EDTA vacutainer tube (Becton Dickinson, USA) for CD4 count, viral load and DR genotyping.

CD4 count and plasma viral load

The CD4/CD4+ cells were enumerated by a bead-based flow cytometry kit (TruCount™, BD Biosciences Immunocytometry Systems, San Jose, CA, USA) and the plasma viral load was estimated using Cobas-Amplicor HIV-1 monitor test, version 1.5 (Roche, Branchburg, NJ) according to manufacturer’s protocol.

Drug resistance genotyping

For DR genotyping, the viral RNA was isolated from plasma using QIAamp viral RNA isolation kit (QIAGEN, Valencia, CA) and reverse transcribed to make complementary DNA (cDNA) copies using a cDNA synthesis kit (MBI Fermentas, Vilnius, Lithuania). The RT region (68 to 231 codon) was amplified using RT specific nested primers as previously described [17]. The PCR products were sequenced in an automated DNA sequencer (ABI PRISM 3730 version 3.0, Applied Biosystems, Foster city, CA, USA). All sequences were subjected to quality assessment and determination of DR mutation profile using the Stanford DR Database (http://hivdb.stanford.edu). NCBI HIV subtyping tool using HIV-1 2009 reference sequence was used to identify the subtype of the virus.

Computational analysis of HIV-1 RT mutations

The computational analysis of the interactions of mutated and wild type RT gene with NRTIs was performed. In this study, the highly identical hits of Crystal structure of HIV-1 RTIs were recovered from Protein Data Bank (PDB). The PDB ID IRTD for RTI with tri-phosphate in the active site of RT with DNA template: primer complex were taken [18]. Molecular models of the wild-type HIV-1 subtype-C RT with NRTI complexes were built from the template. The modeled proteins were optimized by using molecular dynamics (MD) approach together with simulated annealing (SA) methods in the Modeller program [19], and eventually validated on SAVES. The mutated models were created by including reported mutations and optimized. Modeled and mutated HIV-1 subtype C RT and inhibitor complexes were created by docking and binding energy calculations were performed. All drugs were built using SYBYL7.1 (SYBYL7.1, Tripos Inc., St. Louis, MO 63144, USA) and minimized using Tripos force field, GasteigerHuckel, partial atomic charges [20] and Powell's conjugate gradient method with energy gradient convergence criteria of 0.05 kcal/mol [21]. AutoDock Tool was used for preparation of all mutated modeled protein structures for docking. All water molecules were removed from protein structures. Also, polar hydrogen's were added and non polar hydrogen's were merged, finally atom type parameter and Kollman united atom charge was added. For docking study Lamarckian genetic search algorithm was employed and docking run was set to 30. All other parameters were set to default value. Energy calculations were performed by placing the receptor protein inside a user defined 3D grid and allowing various probes to systematically visit each grid point. Each probe consisted of an atom type found in the ligand whose dispersion/repulsion energies and, if appropriate, hydrogen-bonding energies were calculated for all receptor atoms within 8 Å of each grid point. Desolvation energy values were obtained by calculating burial of polar atoms, whilst conformation and torsion energies were calculated according to AMBER force field terms.

Results

Baseline characteristics of the population

All subjects were from north India with age ranging from 7 to 60 years (Median age 32 years) and median CD4 count of 226 cells/µl (Interquartile range: 209.5 to 244.7 cells/µl). More than 92% individuals revealed a history of heterosexual transmission.

The patients who failed the first-line ART had received a combination of 2NRTI+NNRTI treatment for 3–4 years (median time of therapy 42 months; interquartile range 34.3 to 51.0months) and showed clinical signs of treatment failure with baseline median CD4 count of 84.5 cells/µl (interquartilerange 55.5 to 139.4 cells/µl) and viral loads ranging from 1.74E4 to 1.83E6 copies/ml (Table 2).

Subtyping

Of the 122 RT sequences of therapy naïve individuals analysed, 113 (92.6%) clustered around isolate 21068 from India were assigned subtype C; 5 (4.1%) were subtype B (reference HIV-1 isolate BK132 from Thailand), followed by 2 (1.6%) subtype A1 (reference HIV-1 isolate SE7253 from Somalia), and 1 (0.8%) each were 07_BC (reference HIV-1 strain CNGL179 from China) and 08_BC (AF286229 HIV-1 strain 98CN006 from China) HIV-1 circulating recombinant forms.

Pattern of mutations in HIV-1 subtype C reverse transcriptase gene

Therapy naïve: Mutations were defined as amino acid differences from the consensus subtype-B RT sequence (as per Stanford HIV RT sequence database) and polymorphisms were defined when they occurred in at least 1% of the sequences from untreated subjects. Among these, five mutations (121Y (56.3%), 173A (60.17%), 177E (83.18%), 207E (65.45%), 211K (66.36%)) occurred in more than 56% of the HIV-1 isolates and were considered as a local wild type for molecular modeling studies.

In the absence of a subtype-C based database, the amino acid sequence of each isolate in our study was compared to the subtype B consensus amino acid sequence (Stanford HIV RT sequence database) for mutations associated with resistance to RT inhibitors (http://hivdb.stanford.edu). Of the 122 viral isolates from the therapy naïve
HIV-1 individuals, eight showed mutations known to affect the drug susceptibility to RT inhibitors (Table 1). Among these, two isolates had G190A/V, a major NNRTI resistance mutation that is known to cause high level resistance to NVP, intermediate resistance to efavirenz (EFV), and low-level to etrinavir (ETR). Another important mutation, K103N/R detected in HIV isolate from another subject, is also known to cause high-level resistance to NVP (50-fold), DLY (50-fold), and EFV (25-fold). K103R occurred in about 1%-2% of isolates from untreated individuals and by itself has no effect on NNRTI susceptibility. However the combination of K103R with V179D, reduces the susceptibility to each of the NNRTIs about 15-fold. Other mutations detected were: for NNRTI: A98 (V/S) in 4 isolates, K101R in 1, V179D in 3, V179T/I in 6 and for NRTIs: K219Q in 1 isolate, V118I in 2, V75I in 1 and L210F in 1, which are known to be associated with very low-level resistance to certain RTIs.

First-line therapy failures: All thirteen HIV-1 isolates from first-line ART-failure patients had mutations affecting susceptibility to RTI drugs when analyzed using a Stanford DR HIV-1 database (Table 2). The common NNRTI resistance mutations were identified at positions 118 (in 4/13), 184 (in 12/13), 210 (in 6/13) and 215 (in 7/13 isolates) (Figure 1) and common NNRTI resistance mutations were at position 98 (in 5/13), 101 (in 4/13), 103 (in 5/13), 181 (in 5/13) and 190 (in 7/13 isolates) (Figure 2). Our genotyping results from treatment-naïve as well as post-treatment isolates indicate that the PI region is more conserved than the RT region (Figure 3).

### Table 1: Showing drugs affected by mutations detected in eight of 122 HIV-positive ART naïve individual patients in RT-region of pol-gene.

<table>
<thead>
<tr>
<th>Patient</th>
<th>Risk group</th>
<th>CD4/CD3 count (cells/µl)</th>
<th>Viral load (copies/ml)</th>
<th>NRTI</th>
<th>NNRTI</th>
<th>DRUGS AFFECTED</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Blood transfusion</td>
<td>131/830</td>
<td>401,000</td>
<td>V7SI</td>
<td>A98V</td>
<td>Potential low-level resistance: stavudine, didanosine.</td>
</tr>
<tr>
<td>3</td>
<td>Heterosexual</td>
<td>181/1005</td>
<td>29,600</td>
<td>V179D</td>
<td></td>
<td>Potential low-level resistance: delavirdine, efavirenz, etravirine, nevirapine.</td>
</tr>
<tr>
<td>4</td>
<td>Heterosexual</td>
<td>176/1518</td>
<td>81,700</td>
<td>V179D</td>
<td></td>
<td>Potential low-level resistance: delavirdine, efavirenz, etravirine, nevirapine</td>
</tr>
<tr>
<td>5</td>
<td>Heterosexual</td>
<td>59/853</td>
<td>1750,000</td>
<td>V179D</td>
<td></td>
<td>Potential low-level resistance: delavirdine, efavirenz, etravirine, nevirapine</td>
</tr>
<tr>
<td>6</td>
<td>Heterosexual</td>
<td>50/265</td>
<td>1150,000</td>
<td>G190A</td>
<td></td>
<td>Intermediate resistance: efavirenz, Low-level resistance:etravirine, High-level resistance: nevirapine</td>
</tr>
</tbody>
</table>
| 7       | Heterosexual   | 510/3336                  | 11,100                 | K103N|       | High-level resistance:delavirdine, efavirenz, nevirapine, Potential low-level resistance:
| 8       | Heterosexual   | 510/2822                  | 92,200                 | G190V|       | High-level resistance: efavirenz, nevirapine |

### Table 2: Genotypic resistance results in 13 patients on first line of ART treatment with virological failures (HIV-1 RNA> 1000 copies/ml).

<table>
<thead>
<tr>
<th>Patient ID</th>
<th>ART regimen Started</th>
<th>Months</th>
<th>Viral Load</th>
<th>NRTI Resistance Mutations affecting susceptibility</th>
<th>NNRTI Resistance Mutations affecting susceptibility</th>
</tr>
</thead>
<tbody>
<tr>
<td>206</td>
<td>ZDV+3TC+NVP</td>
<td>36</td>
<td>1.24E+05</td>
<td>V75M, V118I, M184V, L210W, T215Y</td>
<td>K101E, Y181C, G190A</td>
</tr>
<tr>
<td>210</td>
<td>D4T+3TC+NVP</td>
<td>46</td>
<td>6.20E+04</td>
<td>F116Y, Q151M, M184V, L210W, T215Y</td>
<td>K103N</td>
</tr>
<tr>
<td>211</td>
<td>ZDV+3TC+EFV</td>
<td>35</td>
<td>9.47E+04</td>
<td>M184V, T215Y</td>
<td>K103N, Y181C, H221Y</td>
</tr>
<tr>
<td>214</td>
<td>D4T+3TC+NVP</td>
<td>25</td>
<td>4.82E+05</td>
<td>L74V, M184V, T215Y</td>
<td>A98G, K101E, Y181C, G190A</td>
</tr>
<tr>
<td>217</td>
<td>D4T+3TC+NVP</td>
<td>45</td>
<td>8.61E+05</td>
<td>M184V, T215Y</td>
<td>A98G, K101E, Y181C, G190A</td>
</tr>
<tr>
<td>218</td>
<td>D4T+3TC+NVP</td>
<td>42</td>
<td>1.92E+05</td>
<td>V118I, M184V, L210S, T215Y</td>
<td>A98G, K101E, Y181C, G190A</td>
</tr>
<tr>
<td>221</td>
<td>D4T30+3TC+NVP</td>
<td>29</td>
<td>4.01E+04</td>
<td>M184V, K219Q</td>
<td>G190A</td>
</tr>
<tr>
<td>222</td>
<td>ZDV+3TC+NVP</td>
<td>51</td>
<td>4.55E+05</td>
<td>M184V, L210W, T215Y</td>
<td>V606A</td>
</tr>
<tr>
<td>223</td>
<td>D4T+3TC+EFV</td>
<td>41</td>
<td>1.83E+06</td>
<td>L74I, M184V, T215F</td>
<td>V901, K103N, V1018, V179F, V181C, H221Y</td>
</tr>
<tr>
<td>228</td>
<td>ZDV+3TC+EFV</td>
<td>82</td>
<td>1.74E+04</td>
<td>M184V</td>
<td>A98G, K103N</td>
</tr>
<tr>
<td>229</td>
<td>D4T+3TC+NVP</td>
<td>42</td>
<td>2.46E+05</td>
<td>V75M, V118I, M184V, L210W, T215Y</td>
<td>A98G, K101E, G190A</td>
</tr>
<tr>
<td>230</td>
<td>D4T+3TC+NVP</td>
<td>45</td>
<td>1.79E+04</td>
<td>M184V, T215Y</td>
<td>K103N, E138Q</td>
</tr>
</tbody>
</table>
therapy naïve isolates was used for molecular modeling of wild type HIV-1 subtype C docked with NRTI using highly identical templates (PDB ID 1RTD). Two isolates from first-line therapy failure patients, SK-206 and SK-208, were selected for the modeling and docking study with difference of just one extra mutation, 75M in SK-206, to see the effect of this mutation in combination with other mutations (118I, 184V, 210W, 215Y) common in both (Figure 4) and compare it with interpretation from Stanford database. Both patients were on ART for 36 months and had high viral load and low CD4 count and revealed mutations affecting drug susceptibility to NRTI as per Stanford DR database (Table 2). On verification and validation of all modeled proteins on SAVES (Structure Analysis and Verification Server) it was observed that all amino acids in the models were located in the same pattern as crystal structures in Ramachandran plots.

On the basis of docking energy scores, we observed a decrease in the binding free energy of 3TC (-22.34 Kcal/mol), ABC (-21.91 Kcal/mol), AZT (-23.61 Kcal/mol), D4T (-23.11 Kcal/mol), DDI (-22.69 Kcal/mol), and FTC (-23.06 Kcal/mol) in case of SK-208 RT model (Table 3). This decrease in the free energy is in concordance with Stanford DR database prediction. On the contrary, our results show the presence of V75M mutation along with other mutations (as found in isolate SK-208) in isolate from SK-206, seem to be partially neutralizing the effect of 118I, 184V, 210W and 215Y mutations in terms of change in the binding energy for drugs D4T (ΔG=-1.46 Kcal/mol), DDI (ΔG=-1.52 Kcal/mol), and FTC (ΔG=-0.57 Kcal/mol) where as in case of AZT there was further increase in binding energy (ΔG=0.87) with the presence of same mutation (75M) which is likely to decrease the susceptibility and might cause high level of resistance (Table 3). On the other hand, in case of isolate SK-206 where Stanford DR database is predicting high to intermediate level resistance to D4T and DDI, our model predict this isolate to be susceptible to these inhibitors due to nullification effect of the drug resistance mutations by V75M, indicating the power of the model to take into consideration the combinatorial effect of all mutations instead of calculating the mutational effect in isolation for each mutation.

Table 3: NRTI mutagenic studies: Docking studies were performed on crystal structure PDB ID 1RTD. All drug molecules were docked with tri-phosphate into the active site of RT complex DNA template-primer. Modes of action of all nucleoside RT inhibitor are inhibition of RT via DNA chain termination after incorporation of the nucleotide analogue. The minimum estimated free energy of binding (Kcal/Mol) at 300K temperature for every docking procedure of RT inhibitor in circulating wild type HIV-1 subtype-C RT and the mutations.
Discussion

Drug resistance is a major obstacle to achieve and maintain virus suppression. In this study, we have sequenced the amplified RT region of pol-gene of HIV isolates from the treatment-naïve as well as first-line ART failure patients in north India and identified drug resistance mutations as per Stanford DR database. There is very high prevalence of DR mutations in the RT gene in HIV-1 isolates from individuals who failed first-line therapy, re-emphasizing the importance of genotyping prior to the start of second line treatment for a rational selection of regimen to have better efficacy of the therapy.

Recent concerns have been raised regarding the first line of ART failures which comprised of RTI drugs. Due to the absence of PI in this regimen in India, a higher rate of DR mutations have been reported in the RT region as compared to PR region in therapy naïve individuals [22].

Among the 4 classes of antiretroviral drugs, the prevalence of resistance in antiretroviral drug-naïve patients is often highest for NNRTIs [23,24]. In the United States, rates of primary NNRTI resistance have been observed to be as high as 16% in certain populations [23]. In our current study isolates from 8 out of 122 therapy naïve individuals had mutations that are known to affect susceptibility to RTIs. Amongst these individuals, 75% (n=6) had mutations affecting NNRTI susceptibility (Table 1). We found high occurrence (2.46%) of 179D mutations compared to the reported frequency of 1% in untreated individuals (http://hivdb.stanford.edu). Mutation V179D/E is known to cause low-level reduction in susceptibility to each of the NNRTIs. The combination of K103R and V179D is indicated to reduce the susceptibility of NVP, DLF, and EFV by 15 folds. The combined effect of this combination on ETR is not known. Other mutations found causing high level resistance to NNRTIs were 190/V/A and K103N.

Consequently the DR genotyping results of isolates from all first line therapy failure patients in our study depicted high level of mutations which are known to cause resistance to RTIs. It was not surprising to find one or more DR mutations in RT region of all the isolates from 13 virological failure individuals (Figure 1 and Figure 2) with G190A being most frequent (53.85%) followed by 103N/A, 98C, 181C (38.46% each) and 101E (30.77%) effecting NNRTI susceptibility while 184V being most frequent (53.85%) followed by 103N/A, 98G, 181C (38.46% each) affecting NNRTI susceptibility (Table 1). The frequency of mutations 118I (30.77%) known to be affecting the susceptibility to NRTIs. It was not surprising to the therapy failure patients in our study depicted high level of mutations in terms of prediction of best 2nd line of therapy for these patients. In order to achieve this we generated a computational NRTI docks HIV-1 subtype C RT model from the consensus sequence of the wild type HIV-1 subtype C in this region, and further verified genotype findings on this model. The computation was verified/validated on SAVE and Ramachandran plot. The RT’ sequences from isolates from two 1st line treatment failure patients were analyzed using the Stanford DR data base as well as our proposed model with constructed crystal RT structure, complexed with DNA template-RNA primer by adding mutations in the wild type subtype-C structure. On the basis of docking energy scores, the effects of mutations on the binding free energy have given very interesting results. In comparison of the binding energy scores obtained after docking of different RTIs with crystal structures of WT sequence or with mutated structure generated after the introduction of mutations detected in isolates SK-208 (V118I, M184V, L210W, T215Y) and SK-206 (V75M, V118I, M184V, L210W, T215Y), the analysis reveals that the presence ofV75M mutation partially neutralizes the resistance effect of mutations 118I, 184V, 210W and 215Y. This may be a combinatorial effect of all mutations occurring together in a particular sequence because of an increase in the number of hydrogen bonds with RTI causing a change in conformation with presence of V75M along with other mutations, as seen in docking attempt of RTIs with structure generated from a sequence of isolated from SK-206 and SK-208, both of which have similar mutations except for the existence of V75M additionally in SK-206 isolate (Table 3). Because of an increase in the number of hydrogen bonds with Lys65 due to presence of V75M mutation, it probably resulted in stabilizing the bound state of RT- NRTI, thus our model predicts patient SK-206 to be susceptible to d4T whereas the Stanford DR database predicts high level of resistance to D4T. On the other hand, in case of AZT, due to the presence of V75M mutation, there is a loss of hydrogen bond in SK-206 as compared to wild type in our model which may culminate to high level resistance to AZT whereas Stanford database analysis suggests intermediate resistance with this mutation alone.

Although the results obtained in our study using the proposed model do overlap to a large extent with the DR interpretation from the Stanford DR database, yet there are a few differences. In this model, we are considering how mutations interact with each other in multiple combinations to affect the susceptibility of the inhibitor which is the actual in vivo situation, rather than their effects in isolation. This justifies the need for a more closely related analysis and a prediction model that is based on individual sequences and co-existence of mutations. Consequently, the homology model proposed here is a modest attempt in that direction and provides a feasibility of predicting the susceptibility or resistance to various drugs and will allow the clinicians to make a rational choice of treatment regimen especially in treatment failure cases. Yet, we believe that the validation of this model coupled with phenotypic studies using a larger number of sequences will reveal its usefulness in a long run.

References


