

Exploring the cause of drug resistance by the detrimental missense mutations in KIT receptor: computational approach

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Abstract In this work, we computationally identified the most detrimental missense mutations of KIT receptor causing gastrointestinal stromal tumors and analyzed the drug resistance of these missense mutations. Out of 31 missense mutations, 19 variants were commonly found less stable, deleterious and damaging by I-Mutant 2.0, SIFT and PolyPhen programs, respectively. Subsequently, we performed modeling of these 19 variants to understand their change in conformations with respect to native KIT receptor by computing their RMSD. Further, the native and 19 mutants were docked with the drug ‘Imatinib’ to explain the drug resistance of these detrimental missense mutations. Among the 19 mutants, we found by docking studies that 12 mutants, namely, F584C, F584L, V654A, L656P, T670I, R804W, D816F, D816V, D816Y, N822K, Y823D and E839K had less binding affinity with Imatinib than the native type. Finally, we analyzed that the loss of binding affinity of these 12 mutants, was due to altered flexibility in their binding amino acids with Imatinib as compared with native type by normal mode analysis. In our work, we found the novel data that the majority of the drug-binding amino acids in those 12 mutants had encountered loss of flexibility, which could be the theoretical basis for the cause of drug insensitivity.

Keywords Missense mutation · KIT receptor · Imatinib · Flexibility

Introduction

Gastrointestinal stromal tumors (GISTs) are the most common mesenchymal tumors of the gastrointestinal tract. These form a distinct category of tumors characterized by oncogenic mutations of the KIT receptor in a majority of patients (Gupta et al. 2008). KIT is a member of the type III family of RTKs, which also includes PDGF-receptor- α and β , CSF-1 receptor (also known as M-CSF receptor or FMS), and the FLT3 receptor (also known as FLK2) (Ullrich and Schlessinger 1990; Blume-Jensen and Hunter 2001). KIT is composed of a glycosylated extracellular ligand-binding domain (ectodomain) that is connected to a cytoplasmic region by means of a single transmembrane (TM) domain (Schlessinger 2000). The ectodomain of KIT and other members of type III RTKs all contain five Ig-like domains, in which the second and third membrane distal domains were shown to play a role in ligand recognition (Ullrich and Schlessinger 1990). The cytoplasmic region of KIT contains a protein tyrosine kinase (PTK) domain with a large kinase-insert region, another hallmark of type III RTKs. Binding of SCF to KIT leads to receptor dimerization, intermolecular autophosphorylation, and PTK activation. It was proposed that the fourth Ig-like domain of KIT is responsible for KIT dimerization in response to either monovalent or bivalent SCF binding (Lev et al. 1992; Blechman et al. 1995). However, other studies have demonstrated that ligand-induced dimerization of KIT is driven by bivalent binding of SCF (Philo et al. 1996; Lemmon et al. 1997).

KIT is used not only for diagnosis but also for targeted therapy of GISTs. Imatinib (also known as Gleevec or

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STI571), a tyrosine kinase inhibitor, is widely used in the treatment of advanced and metastatic GISTs. Specific mutations in the KIT gene result in GISTs that are relatively unresponsive to the Imatinib treatment (Bauer et al. 2005; Jonathan Fletcher and Brian Rubin 2007; Engelman and Settleman 2008). Identifying the disease-associated missense mutation and drug resistance caused by them is a challenging task for cancer research. So, we have given an attempt to investigate the KIT receptor by our devised computational protocol which we established, implemented and analyzed with HER2 and CDKN2A (Rajasekaran et al. 2008a, b). With this continuation, we undertook and implemented our devised protocol to identify the detrimental missense mutations in KIT protein and proposed a modeled structure for the mutants. Then we docked the inhibitor, Imatinib with both the native and mutants of KIT receptor to find out the binding effect and the nature of flexibility in the binding pockets to understand the cause of drug resistance by these missense mutations.

Materials and methods

Data sets

The protein sequence and variants (single amino acid polymorphisms/missense mutations/point mutants) for KIT receptor were obtained from the Swissprot, available at <http://www.expasy.ch/sprot/> to find out the detrimental point mutants. The subsection of each Swissprot entry provides information on polymorphic variants, in which some polymorphic variants may be disease(s) associated with defects in a given protein; most of them are nsSNPs (non-synonymous SNPs) in gene sequence and SAPs (single amino acid polymorphisms) in protein sequence (Yip et al. 2004, 2008; Boeckmann et al. 2003). The 3D Cartesian coordinates of KIT receptor and KIT-drug complex were obtained from Protein Data Bank (Berman et al. 2000) for *in silico* mutation modeling and docking studies based on detrimental missense mutations.

Predicting stability change on single amino acid polymorphism based on support vector machine (I-Mutant 2.0)

We used the program I-Mutant 2.0 available at <http://gpcr.biocomp.unibo.it/cgi/predictors/I-Mutant2.0/I-Mutant2.0.cgi>. The I-Mutant 2.0 is a support vector machine (SVM)-based tool for the automatic prediction of protein stability changes upon single-point mutations; I-Mutant 2.0 predictions are performed starting either from the protein structure or, more importantly, from the protein sequence (Capriotti et al. 2005). This program was trained and tested on a

data set derived from ProTherm (Bava et al. 2004), which is presently the most comprehensive available database of thermodynamic experimental data of free energy changes of protein stability upon mutation under different conditions. The output file shows the predicted free energy change value or sign ($\Delta\Delta G$), which is calculated from the unfolding Gibbs free energy value of the mutated protein minus the unfolding Gibbs free energy value of the native type (kcal/mol). Positive $\Delta\Delta G$ values mean that the mutated protein possesses high stability and vice versa.

Analysis of functional consequences of point mutant by sequence homology based method (SIFT)

We used the program SIFT (Ng and Henikoff 2003), which is specifically to detect the deleterious single amino acid polymorphism, available at <http://blocks.fhcrc.org/sift/SIFT.html>. SIFT is a sequence homology based tool, which presumes that important amino acids will be conserved in the protein family. Hence, changes at well-conserved positions tend to be predicted as deleterious (Ng and Henikoff 2003). We submitted the query in the form of protein sequences. The underlying principle of this program is that, SIFT takes a query sequence and uses multiple alignment information to predict tolerated and deleterious substitutions for every position of the query sequence. SIFT is a multistep procedure that, given a protein sequence, (a) searches for similar sequences, (b) chooses closely related sequences that may share similar function, (c) obtains the multiple alignment of these chosen sequences, and (d) calculates normalized probabilities for all possible substitutions at each position from the alignment. Substitutions at each position with normalized probabilities less than a chosen cutoff are predicted to be deleterious and greater than or equal to the cutoff are predicted to be tolerated (Ng and Henikoff 2001). The cutoff value in SIFT program is tolerance index of ≥ 0.05 . Higher the tolerance index, less functional impact a particular amino acid substitution is likely to have.

Simulation for functional change in point mutant by structure homology based method (PolyPhen)

Analyzing the damaged point mutations at the structural level is considered to be very important to understand the functional activity of the concerned protein. We used the server PolyPhen (Ramensky et al. 2002) which is available at <http://coot.embl.de/PolyPhen/> for this purpose. The input options for the PolyPhen server are protein sequence or SWALL database ID or accession number together with sequence position with two amino acid variants. We submitted the query in the form of protein sequence with

mutational position and two amino acid variants. Sequence-based characterization of the substitution site, profile analysis of homologous sequences and mapping of the substitution site to known protein three-dimensional structures are the parameters taken into account by the PolyPhen server to calculate the score. It calculates position-specific independent count (PSIC) scores for each of the two variants, and then computes the PSIC scores' difference between them. Higher the PSIC score difference, higher is the functional impact a particular amino acid substitution is likely to have.

Modeling the effect of mutations on protein flexibility

Structure analysis was performed for evaluating the structural deviation between native type and mutant types by means of RMSD (root mean square deviation). We used the web resource Protein Data Bank (Berman et al. 2000) and single amino acid polymorphism database (SAAPdb) (Cavallo and Martin 2005) to identify the 3D structure of KIT receptor (PDB ID: 1t45). We also confirmed the mutation position and the mutation residue in PDB ID: 1t45. The mutation was performed by using SWISSPDB viewer and energy minimization for 3D structures was performed by NOMAD-Ref server (Lindahl et al. 2006). This server use Gromacs as default force field for energy minimization based on the methods of steepest descent, conjugate gradient and L-BFGS methods (Delarue and Dumas 2004). We used the conjugate gradient method to minimize the energy of the 3D structure of KIT protein. In order to optimize the 3D structure of KIT protein, we used the ifold server (Sharma et al. 2006) for simulated annealing, which is based on discrete molecular dynamics and is one of the fastest strategies for simulating protein dynamics. This server is also efficient in sampling the vast conformation space of biomolecules in both length and time scales. Divergence in mutant structure with native structure is due to mutations, deletions, and insertions (Han et al. 2006) and the deviation between the two structures could alter the functional activity (Varfolomeev et al. 2002) with respect to binding efficiency of the inhibitor which is evaluated by their RMSD values.

Identification of binding sites and computation of atomic contact energy (ACE) between KIT receptor and inhibitor

In order to compute the atomic contact energy (ACE) between KIT receptor and with inhibitor, we submitted the PDB ID: 1t46, a complex of KIT receptor with Imatinib, into the ligand contact tool (LCT) program available at <http://firedb.bioinfo.cnio.es/Php/Contact.php> (López et al. 2007). This server calculates contacts between the binding

residues of KIT receptor with Imatinib with default parameters. Then, we unbound the Imatinib from the KIT receptor of the PDB ID: 1t46 to perform point mutations on KIT protein by the SWISSPDB viewer, energy minimization by NomadRef and simulated annealing by ifold. Finally, we used the program PatchDock for docking the native and mutants of KIT receptor with the Imatinib to compute ACE by using additional option of binding residue parameter. The underlying principle of this server is based on molecular shape representation, surface patch matching plus filtering and scoring (Duhovny et al. 2002). It is aimed at finding docking transformations that yield good molecular shape complementarity. Such transformations, when applied, induce both wide interface areas and small amounts of steric clashes. A wide interface is ensured to include several matched local features of the docked molecules that have complementary characteristics. The PatchDock algorithm divides the Connolly dot surface representation (Connolly 1983a, b) of the molecules into concave, convex and flat patches. Then, complementary patches are matched in order to generate candidate transformations. Each candidate transformation is further evaluated by a scoring function that considers both geometric fit and atomic desolvation energy (Schneidman-Duhovny et al. 2005; Zhang et al. 1997). Finally, an RMSD (root mean square deviation) clustering is applied to the candidate solutions to discard redundant solutions. The main reason behind PatchDock's high efficiency is its fast transformational search, which is driven by local feature matching rather than brute force searching of the six-dimensional transformation space. It further speeds up the computational processing time by utilizing advanced data structures and spatial pattern detection techniques, such as geometric hashing and pose clustering.

Exploring the flexibility of binding pocket by normal mode analysis

A quantitative measure of the atomic motions in proteins can be obtained from the mean square fluctuations of the atoms relative to their average positions. These can be related to the *B*-factor (Yuan et al. 2005; Ringe and Petsko 1986). Analysis of *B*-factors, therefore, is likely to provide newer insights into protein dynamics, flexibility of amino acids, and protein stability (Parthasarathy and Murthy 2000). It is to be noted that protein flexibility is important for protein function and for rational drug design (Carlson and McCammon 2000). Also flexibility of certain amino acids in protein is useful for various types of interactions. Moreover, flexibility of amino acids in drug binding pocket is considered to be a significant parameter to understand the binding efficiency. In fact, loss of flexibility impairs the binding effect (Hinkle and Tobacman 2003) and vice versa

(Rajasekaran et al. 2008a). Hence, this can be analyzed by the *B*-factor, which is computed from the mean-square displacement $\langle R^2 \rangle$ of the lowest-frequency normal mode using the ElNémo server (Suhre and Sanejouand 2004).

Results and discussion

Single amino acid polymorphism data set from Swissprot

The KIT receptor and a total of 31 variants, namely, V532I, M541L, K550I, V559A, V559D, E583K, F584C, F584L, G601R, V654A, L656P, G664R, T670I, C691S, S715N, D737N, R791G, R796G, R804W, G812V, D816F, D816H, D816V, D816Y, D820G, D820Y, N822K, Y823D, A829P, E839K and T847P investigated in this work were retrieved from the Swissprot database (Yip et al. 2004, 2008; Boeckmann et al. 2003).

Identification of functional variants by I-Mutant 2.0

Out of 31 variants, we obtained 29 variants found to be less stable from the I-Mutant 2.0 server (Capriotti et al. 2005) as shown in Table 1. Out of such 29 variants, 7 variants, viz., F584C, F584L, G664R, D737N, G812V, D816V and D820G showed a $\Delta\Delta G$ value of > -2.0 . The other 15 variants, viz., V532I, M541L, E583K, G601R, V654A, L656P, S715N, D816F, D816H, D816Y, D820Y, N822K, A829P, E839K and T847P showed a $\Delta\Delta G$ value of > -1.0 and the remaining 7 variants showed a $\Delta\Delta G$ value of < -1.0 as depicted in Table 1.

Out of 29 variants which showed negative $\Delta\Delta G$, 4 variants, namely, E583K, G664R, D816H and E839K changed their amino acid from negatively charged to positively charged amino acid; 2 variants, L656P and A829P changed from non-polar to polar; 2 variants, S715N and N822K changed from polar to positively charged; 2 variants, R791G and R796G from positively charged to non-polar; 2 variants, D816V and D820G changed from negatively charged to non-polar; 2 variants, D816Y and D820Y changed from negatively charged to aromatic, followed by 1 variant each, viz., F584C, F584L, G601R, T670I, D737N, R804W, D816F and Y823D changed from aromatic to polar, aromatic to non-polar, non-polar to positively charged, polar to non-polar, negatively charged to polar, positively charged to aromatic, non-polar to aromatic and aromatic amino acid to negatively charged amino acid, respectively. It is also to be noted that five other variants, V532I, M541L, V559A, V654A and G812V which retained their non-polar properties, and one other variant, T847P which retained its polar property were found to be less stable by I-Mutant 2.0. The above point

Table 1 List of variants that were predicted to be functionally significant by I-Mutant 2.0, SIFT and PolyPhen

AA change	$\Delta\Delta G$	Tolerance index	PSIC SD
V532I	-1.10	1.00	0.115
M541L	-1.03	0.56	0.563
K550I	0.45	0.07	3.168
V559A	-0.70	0.92	2.094
V559D	-0.80	0.34	2.686
E583K	-1.62	0.00	2.256
F584C	-2.17	0.00	2.853
F584L	-2.53	0.05	1.667
G601R	-1.60	0.00	2.790
V654A	-1.56	0.02	1.681
L656P	-1.54	0.00	2.871
G664R	-2.14	0.25	2.790
T670I	-0.27	0.02	2.692
C691S	0.07	0.79	1.176
S715N	-1.29	0.57	1.061
D737N	-3.31	0.13	1.089
R791G	-0.43	0.55	3.136
R796G	-0.18	0.00	3.136
R804W	-0.79	0.03	2.823
G812V	-3.42	0.00	3.015
D816F	-1.43	0.00	3.130
D816H	-1.91	0.02	1.132
D816V	-2.08	0.00	3.185
D816Y	-1.56	0.00	2.995
D820G	-2.82	0.27	2.789
D820Y	-1.69	0.01	3.239
N822K	-1.44	0.04	2.569
Y823D	-0.39	0.00	2.307
A829P	-1.44	0.02	1.983
E839K	-1.39	0.00	2.256
T847P	-1.03	0.00	2.557

AA change amino acid change

Letters in bold indicate less stable, deleterious and damaging by I-Mutant, SIFT and PolyPhen, respectively

portrays that preserving amino acid physico-chemical properties does not necessary result in harmless mutation. Indeed considering only amino acid substitution based on physico-chemical properties could not be able to identify the detrimental effect rather than considering the sequence conservation along with the above said properties could have more advantages and reliable to find out the detrimental effect of missense mutations (Teng et al. 2009).

Deleterious single point mutant by SIFT program

Protein sequences of 31 variants were submitted independently to the SIFT program (Ng and Henikoff 2003) to

check its tolerance index. Among the 31 variants, 20 variants were found to be deleterious having the tolerance index score of ≤ 0.05 . The results are shown in Table 1. We observed that out of 20 such deleterious variants, 12 variants, E583K, F584C, G601R, L656P, R796G, G812V, D816F, D816V, D816Y, Y823D, E839K and T847P showed a highly deleterious tolerance index score of 0.00, 1 variant, D820Y with a tolerance index score of 0.01, 4 variants, V654A, T670I, D816H and A829P with a tolerance index score of 0.02, 1 variant, R804W with a tolerance index score of 0.03, 1 variant, N822K with a tolerance index score of 0.04 and 1 variant, F584L with a tolerance index score of 0.05. It was interesting to observe that all the deleterious variants according to SIFT also were seen to be less stable by the I-Mutant 2.0 server.

Damaged single point mutant by PolyPhen server

Protein sequence with mutational position and amino acid variants associated with 31 single point mutants, investigated in this work were submitted as inputs to the PolyPhen server (Ramensky et al. 2002) and the results are shown in Table 1. A PSIC score difference of 1.5 and above is considered to be damaging. It can be seen that, out of 31 variants, 25 variants were considered to be damaging by PolyPhen. Also these 25 variants exhibited a PSIC score difference between 1.667 and 3.239. It is to be noted that all the variants except K550I which were considered to be damaging by PolyPhen also were seen to be less stable according to the I-Mutant 2.0 server. Similarly, all the variants except K550I, V559A, V559D, G664R, R791G and D820G which were considered to be damaging by PolyPhen also were seen to be deleterious according to the SIFT server.

Rational consideration of detrimental point mutations

We rationally considered the most 19 potential detrimental point mutations for further course of investigations which were E583K, F584C, F584L, G601R, V654A, L656P, T670I, R796G, R804W, G812V, D816F, D816V, D816Y, D820Y, N822K, Y823D, A829P, E839K and T847P as they were commonly found less stable, deleterious and damaging by the I-Mutant 2.0, SIFT and PolyPhen servers, respectively. Among these, 9 variants, V654A, T670I, G812V, D816V, D820Y, N822K, Y823D, A829P and E839K showed very good agreement with experimental observation performed elsewhere (Kim et al. 2005; Nishida et al. 2008; Ciccarelli et al. 2000; Ma et al. 2002; Weisberg and Griffin 2003; Fletcher and Rubin 2007; Kemmer et al. 2004; Chen et al. 2005; Roberts et al. 2007; McLean et al. 2005; Chen et al. 2004; Growney et al. 2005; Gajiwala et al. 2009; Debiec-Rychter et al. 2005).

Computing RMSD by modeling of mutant structures

Mapping the 19 potential detrimental variants into protein structure information was obtained from SAAPdb (Cavallo and Martin 2005). Mutation at specified position was performed by SWISSPDB viewer independently to get modeled structures. Then, energy minimizations and simulated annealing were performed by NOMAD-Ref server (Lindahl et al. 2006) and ifold server (Sharma et al. 2006), respectively, for both the native structure (PDB 1t45) and mutant modeled structures (E583K, F584C, F584L, G601R, V654A, L656P, T670I, R796G, R804W, G812V, D816F, D816V, D816Y, D820Y, N822K, Y823D, A829P, E839K and T847P).

In order to find out the deviation between the two structures, we superimposed the energy refined native structure (PDB 1t45) with all the energy refined mutant modeled structures to get RMSD. The higher the RMSD value, the more is the deviation between the native and the mutant structure, which in turn changes their binding efficiency with inhibitors due to deviation in the 3D space of the binding residues of KIT receptor. Table 2 shows the RMSD for native structure with all the mutant modeled structures. It can be seen from Table 2 that, four mutants, V654A, T670I, Y823D and E839K exhibited a high RMSD value >3.00 Å; two mutants, D816F and D816V with RMSD >2.00 Å; six mutants, F584C, F584L, L656P,

Table 2 RMSD and ACE for native and mutants

KIT receptor	RMSD (Å)	ACE (Kcal/mol)
Native	0.00	−261.73
E583K	0.06	−293.98
F584C	1.69	−207.98
F584L	1.33	−213.86
G601R	0.91	−276.75
V654A	3.11	−147.43
L656P	1.41	−224.34
T670I	3.69	−121.48
R796G	0.98	−279.71
R804W	1.32	−230.78
G812V	0.79	−299.67
D816F	2.57	−167.43
D816V	2.92	−137.15
D816Y	1.67	−246.58
D820Y	0.60	−290.19
N822K	1.45	−233.18
Y823D	3.14	−198.85
A829P	0.31	−310.64
E839K	3.06	−113.31
T847P	0.68	−267.40

Letters in bold indicate mutants with less binding affinity than native

R804W, D816Y and N822K with RMSD >1.00 Å; the remaining seven variants exhibited a low RMSD value ≤ 0.98 Å. Figure 1a shows the superimposed structure of native KIT with the mutant D816V having the RMSD of 2.92 Å as an illustrative example of this work.

Rationale of binding efficiency for native and mutant KIT receptor with its inhibitor

In order to understand the binding efficiency of KIT protein with its inhibitor, we selected the PDB ID: 1t46 which has KIT complex with Imatinib. The LCT program was used to calculate contacts between the binding residues of KIT and inhibitor. It can be seen from Table 3 that a total of 29 amino acids, viz., Leu(595), Val(603), Ala(621), Val(622), Lys(623), Glu(640), Val(643), Leu(644), Leu(647), Ile(653), Val(654), Val(668), Ile(669), Thr(670), Glu(671), Tyr(672), Cys(673), Gly(676), Asp(677), Leu(783), Cys(788), Ile(789), His(790), Arg(791), Leu(799), Ile(808), Cys(809), Asp(810) and Phe(811) act as binding residues for KIT complex with Imatinib. It was interesting to observe that some of these amino acids include Lys(623), Glu(640), Thr(670), Cys(673), Asp(810) and Phe(811) that act as key binding residues of KIT with Imatinib which was also proved experimentally by other groups (Mol et al. 2004; Roskoski 2005). To understand the binding efficiency of Imatinib with both native and mutants of KIT structure, we unbound the Imatinib from KIT from the PDB ID: 1t46. We once again performed the point mutation by SWISSPDB viewer for the 19 variants. Energy minimization was performed by Nomad-Ref (Lindahl et al. 2006) followed by simulated annealing with ifold server (Sharma et al. 2006) to get optimized structures for both native and mutants.

Docking was performed using the PatchDock server between Imatinib with both native and mutant modeled structures of KIT to find out the binding efficiency in the form of atomic contact energy (ACE) as can be seen from

Table 2. By this analysis, the ACE between Imatinib and native KIT was found to be -261.73 , whereas with mutants, the ACE was found to be between the ranges -113.31 and -310.64 . It can be seen from Table 2 that 7 mutants, E583K, G601R, R796G, G812V, D820Y, A829P and T847P established high binding affinity with Imatinib than native KIT; the remaining 12 mutants were found to have less binding affinity with Imatinib than native KIT in terms of ACE. These data clearly portray that Imatinib is considered to be the most favorable inhibitor not only for native but also for the former seven mutants. Also we emphasize that the maximum binding effect of Imatinib with these mutants might be due to the 3D conformation of Imatinib which exclusively made comfortable fit with high ACE into the 3D space of the binding residues of these former mutants as compared to the later mutants. But the aim of our study was to find out the most detrimental point mutants causing drug resistance. Hence the later 12 mutants, F584C, F584L, V654A, L656P, T670I, R804W, D816F, D816Y, D816Y, N822K, Y823D and E839K, which had less binding affinity with Imatinib in terms of ACE, as they have the RMSD of about ≥ 1.32 to ≤ 3.69 Å, were considered to be rationally significant missense mutations with respect to drug resistance as can be seen from Table 2. This analysis reveals that the some of these mutants, viz., V654A, T670I, D816V, N822K, Y823D and E839K that render resistance with Imatinib are also in good agreement with clinical observations (Ma et al. 2002; Weisberg and Griffin 2003; Fletcher and Rubin 2007; Kemmer et al. 2004; Chen et al. 2005; Roberts et al. 2007; McLean et al. 2005; Chen et al. 2004; Growney et al. 2005; Gajiwala et al. 2009; Debiec-Rychter et al. 2005; Tan et al. 2006; Guo et al. 2007). Figure 1b shows the superimposed structure of binding region of native complex (i.e., native KIT receptor and Imatinib) with mutant D816V complex (i.e., mutant D816V and Imatinib) as an another illustrative example of this work.

Fig. 1 a Superimposed structure of native with mutant D816V **b** Superimposed structure of binding region of native complex (native KIT receptor and imatinib) with mutant D816V complex (mutant D816V and imatinib)

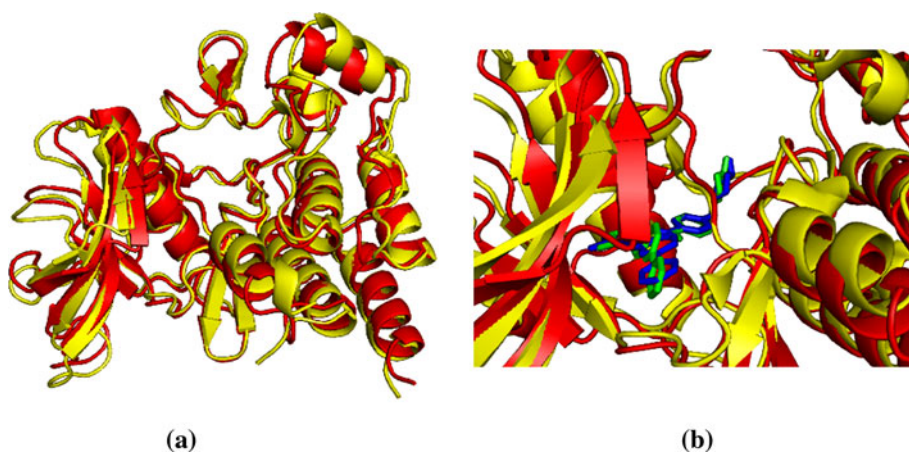


Table 3 Comparison of normalized mean square displacement of drug binding amino acids of native and mutants

Binding residues	Normalized mean square displacement $\langle R^2 \rangle$												
	Native	F584C	F584L	V654A	L656P	T670I	R804W	D816F	D816V	D816Y	N822K	Y823D	E839K
Leu(595)	0.0070	0.0067*	0.0071	0.0064*	0.0070	0.0070	0.0068*	0.0075	0.0055*	0.0069*	0.0065*	0.0021*	0.0066*
Val(603)	0.0037	0.0037	0.0039	0.0036*	0.0036*	0.0035*	0.0037	0.0042	0.0022*	0.0038	0.0034*	0.0047	0.0036*
Ala(621)	0.0041	0.0040*	0.0044	0.0041	0.0044	0.0044	0.0040*	0.0043	0.0039*	0.0043	0.0040*	0.0032*	0.0042
Val(622)	0.0012	0.0014	0.0014	0.0019	0.0015	0.0016	0.0012	0.0013	0.0014	0.0016	0.0013	0.0060	0.0017
Lys(623)	0.0009	0.0014	0.0009	0.0015	0.0009	0.0010	0.0010	0.0016	0.0021	0.0013	0.0009	0.0060	0.0013
Glu(640)	0.0026	0.0037	0.0026	0.0040	0.0025*	0.0032	0.0026	0.0040	0.0047	0.0047	0.0027	0.0157	0.0032
Val(643)	0.0009	0.0030	0.0005*	0.0040	0.0013	0.0010	0.0016	0.0027	0.0019	0.0029	0.0016	0.0151	0.0033
Leu(644)	0.0015	0.0032	0.0011*	0.0036	0.0020	0.0012*	0.0021	0.0023	0.0011*	0.0023	0.0020	0.0115	0.0034
Leu(647)	0.0027	0.0036	0.0024*	0.0045	0.0031	0.0024*	0.0030	0.0027	0.0022*	0.0023*	0.0030	0.0147	0.0041
Ile (653)	0.0043	0.0037*	0.0042*	0.0035*	0.0042*	0.0038*	0.0040*	0.0037*	0.0041*	0.0042*	0.0042*	0.0093	0.0037*
Val(654)	0.0035	0.0031*	0.0035	0.0032*	0.0035	0.0033*	0.0034*	0.0029*	0.0032*	0.0032*	0.0036	0.0080	0.0035
Val(668)	0.0038	0.0043	0.0035*	0.0029*	0.0038	0.0037*	0.0035*	0.0040	0.0038	0.0036*	0.0034*	0.0063	0.0033*
Ile (669)	0.0045	0.0045	0.0043*	0.0032*	0.0046	0.0042*	0.0042*	0.0044*	0.0040*	0.0046	0.0042*	0.0048	0.0039*
Thr(670)	0.0043	0.0046	0.0038*	0.0040*	0.0047	0.0043	0.0044	0.0042*	0.0042*	0.0041*	0.0043	0.0036*	0.0045
Glu(671)	0.0053	0.0054	0.0050*	0.0052*	0.0055	0.0050*	0.0053	0.0052*	0.0048*	0.0049*	0.0053	0.0046*	0.0057
Tyr(672)	0.0066	0.0065*	0.0065*	0.0063*	0.0068	0.0062*	0.0065*	0.0067	0.0061*	0.0063*	0.0062*	0.0054*	0.0067
Cys(673)	0.0065	0.0062*	0.0063*	0.0060*	0.0068	0.0058*	0.0063*	0.0065	0.0051*	0.0063*	0.0063*	0.0052*	0.0065
Gly(676)	0.0071	0.0067*	0.0072	0.0061*	0.0074	0.0067*	0.0070*	0.0071	0.0062*	0.0070	0.0070*	0.0092	0.0070*
Asp(677)	0.0071	0.0064*	0.0070*	0.0053*	0.0070*	0.0065*	0.0066*	0.0065*	0.0071	0.0065*	0.0067*	0.0069*	0.0063*
Leu(783)	0.0053	0.0044*	0.0051*	0.0031*	0.0048*	0.0049*	0.0046*	0.0049*	0.0049*	0.0052*	0.0047*	0.0050*	0.0036*
Cys(788)	0.0045	0.0033*	0.0045	0.0017*	0.0040*	0.0041*	0.0037*	0.0039*	0.0034*	0.0042*	0.0038*	0.0041*	0.0023*
Ile (789)	0.0057	0.0049*	0.0055*	0.0026*	0.0052*	0.0053*	0.0050*	0.0055*	0.0051*	0.0056*	0.0050*	0.0049*	0.0036*
His(790)	0.0055	0.0047*	0.0053*	0.0027*	0.0050*	0.0051*	0.0048*	0.0051*	0.0053*	0.0055	0.0049*	0.0051*	0.0037*
Arg(791)	0.0057	0.0050*	0.0056*	0.0028*	0.0052*	0.0054*	0.0051*	0.0055*	0.0052*	0.0059	0.0051*	0.0047*	0.0038*
Leu(799)	0.0058	0.0052*	0.0058	0.0046*	0.0059	0.0053*	0.0055*	0.0054*	0.0043*	0.0055*	0.0056*	0.0052*	0.0054*
Ile (808)	0.0049	0.0040*	0.0048*	0.0034*	0.0047*	0.0043*	0.0044*	0.0043*	0.0042*	0.0047*	0.0046*	0.0031*	0.0040*
Cys(809)	0.0042	0.0033*	0.0042	0.0025*	0.0039*	0.0038*	0.0037*	0.0037*	0.0041*	0.0042	0.0038*	0.0039*	0.0032*
Asp(810)	0.0041	0.0032*	0.0042	0.0018*	0.0037*	0.0038*	0.0035*	0.0037*	0.0040*	0.0044	0.0036*	0.0027*	0.0027*
Phe(811)	0.0045	0.0037*	0.0046	0.0022*	0.0040*	0.0044*	0.0040*	0.0047	0.0060	0.0050	0.0040*	0.0041*	0.0030*

Bold number indicates amino acids with identical flexibility of both native and mutants

* Indicates amino acids with decreased flexibility of mutants than native

Majority of amino acids in drug binding pocket show loss of flexibility

In order to understand the cause of drug insensitivity by these 12 detrimental missense mutations, we used the program Elnemo (Suhre and Sanejouand 2004) to compare the flexibility of amino acids of both native and mutants, which are involved in binding with Imatinib. Table 3 depicts the flexibility of amino acids in drug binding pocket of both native and mutants by means of normalized mean-square displacement $\langle R^2 \rangle$. We further sorted out these data into three different ranges of flexibility. One is the $\langle R^2 \rangle$ of amino acids in drug binding pocket of mutants which is exactly the same as $\langle R^2 \rangle$ of the amino acids in drug binding pocket of native named as 'identical

flexibility'. The other is the $\langle R^2 \rangle$ of amino acids in drug binding pocket of mutants which is higher than $\langle R^2 \rangle$ of the amino acids in drug binding pocket of native named as 'increased flexibility'. And the last is the $\langle R^2 \rangle$ of amino acids in drug binding pocket of mutants which is lesser than $\langle R^2 \rangle$ of amino acids in drug binding pocket of native named as 'decreased flexibility'. Table 4 depicted that the three different ranges of flexibility of amino acids in drug binding pocket for 12 mutants. From this analysis, we found that the number of drug binding amino acids of these 12 mutants having the range of identical flexibility is lesser than those of increased and decreased flexibility. On the other hand the number of drug binding amino acids having the range of increased flexibility is lesser than those of decreased flexibility as could be seen from Table 4. This

Table 4 Drug binding amino acids of mutants with different ranges of flexibility based on $\langle R2 \rangle$

Mutants	A $\langle R2 \rangle$	B	C
F584C	18	2	9
F584L	16	6	7
V654A	22	1	6
L656P	13	4	12
T670I	22	2	5
R804W	20	4	5
D816F	15	3	11
D816V	22	2	5
D816Y	15	2	12
N822K	20	3	6
Y823D	17	0	12
E839K	17	2	10

Where the letter 'A' denotes amino acids with decreased flexibility of mutants than native; 'B' denotes amino acids with identical flexibility of both native and mutants; 'C' denotes amino acids with increased flexibility of both native and mutants

evidently exemplified that majority of amino acids participated in drug binding pocket of these mutants lost their flexibility due to their occurrence in the range of 'decreased flexibility' which signify the loss of binding efficiency with the inhibitor, Imatinib. The supplementary tables are given for all the 31 mutants with RMSD, ACE, flexibility of binding amino acids and possible citations for drug resistant mutants to support our work.

Conclusion

As several missense mutations like V560G, D820A and K641E of Kit receptor showed drug resistance (Fletcher and Rubin 2007; Heinrich et al. 2006; Bauer et al. 2006), in this work, we also found quite a few drug-resistant mutations by our computational approach. Based on our investigation, we concluded that out of 31 variants which were retrieved from Swissprot, 29 variants were found less stable by I-Mutant 2.0; 20 variants were found deleterious by SIFT and 25 variants were considered damaging by PolyPhen. We rationally considered that the 19 variants were considered to be the potential detrimental point mutations as they were commonly found to be less stable, deleterious and damaging by the I-Mutant 2.0, SIFT and PolyPhen servers, respectively. Further we modeled the mutants based on these 19 variants. The RMSD for mutants with native was computed which showed the range between 0.06 and 3.69 Å. Moreover docking was performed between Imatinib with native and mutants which showed the ACE between -113.31 and -310.64. Finally we

conclude that the less binding affinity of 12 mutants, viz., F584C, F584L, V654A, L656P, T670I, R804W, D816F, D816Y, D816V, N822K, Y823D and E839K with Imatinib when compared with native KIT structure in terms of ACE, as they have the RMSD between ≥ 1.32 and ≤ 3.69 Å, were persuasively considered to be significant for the cause of drug resistance. We also elucidated the cause of drug resistance of these 12 mutants in terms of normalized mean square displacement $\langle R2 \rangle$ by normal mode analysis. From this analysis, we concluded that the majority of amino acids of these mutants participated as binding residues with Imatinib had 'decreased flexibility' which could be the cause for drug insensitivity. The overall scope of our result is (1) to consider computationally a suitable protocol for missense mutation (point mutation/single amino acid polymorphism) analysis before wet lab experimentation, (2) to provide an optimal path for further clinical and experimental studies to characterize KIT mutants in depth, (3) to consider an appropriate protocol for identifying the drug resistance by point mutations, (4) to validate the potency of drug molecules and (5) to select the potential drug based on point mutation associated cancer targets.

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