

### Identification of hot spot residues at protein-protein interface

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#### Abstract:

It is known that binding free energy of protein-protein interaction is mainly contributed by hot spot (high energy) interface residues. Here, we investigate the characteristics of hot spots by examining inter-atomic sidechain-sidechain interactions using a dataset of 296 alanine-mutated interface residues. Results show that hot spots participate in strong and energetically favorable sidechain-sidechain interactions. Subsequently, we describe a novel, yet simple 'hot spot' prediction model with an accuracy that is similar to many available approaches. The model is also shown to efficiently distinguish specific protein-protein interactions from non-specific interactions.

**Key words:** protein-protein interaction; interface analysis; hot spot residues; inter-atomic interaction

#### Abbreviations:

BPTI = bovine pancreatic trypsin inhibitor; NA = number of atoms in a residue involving sidechain-sidechain interactions at the interface;  $NA_l$  = number of atoms with legitimate (favorable) sidechain-sidechain contacts;  $NA_{il}$  = number of atoms with illegitimate (unfavorable) sidechain-sidechain contacts;  $NC_l$  = number of legitimate (favorable) sidechain-sidechain contacts;  $NC_{il}$  = number of illegitimate (unfavorable) sidechain-sidechain contacts; PDB = protein databank; vdW = van der Waals; NPV = native predictive value; PPV = positive predictive value; SN = sensitivity; SP = specificity

#### Background:

Many biological processes such as signal transduction, transport, cellular motion and regulatory mechanisms are mediated by protein-protein interactions. The study of protein-protein interactions has gained momentum for deciphering the specificity of protein-protein interfaces. Many parameters (e.g. interface hydrophobicity, residue frequencies and pairing preferences at interface) have been defined to describe interface features. [1, 2, 3, 4, 5, 6, 7] Recently, the contribution of individual residues to subunit interactions have been estimated using alanine-scanning mutagenesis, where the mutation of a target residue to alanine is followed by the measure of  $\Delta\Delta G$  (change in binding free energies), as described elsewhere. [8] The binding free energy is observed to be dominantly contributed by high energy residues, called 'hot spots'. [9, 10] For example, at the BPTI-trypsin interface, hot spot Lys15->Ala mutation ( $\Delta\Delta G = 10 \text{ kcal}\cdot\text{mol}^{-1}$ ) leads to a 200-fold decrease in association rate, while low energy residue ARG17->ALA ( $\Delta\Delta G < 0.5 \text{ kcal}\cdot\text{mol}^{-1}$ ) has little effect on association rate. [11] Therefore, interface specificity is effectively determined by hot spots.

Because hot spots are a good indicator of interface specificity, their characteristics have been widely investigated. [10, 12, 13, 14, 15, 16, 17, 18] Hot spots are enriched in TRP, TYR and ARG and are often surrounded by hydrophobic rings to occlude bulk solvent. [10] In addition, hot spots statistically correlate with structurally conserved residues in ten protein families. [12] Moreover, hot spots from different monomers prefer to interact and their couplings are structurally conserved. [13] It has also been found that hot spots are related to central interface residues using the small-world network approach (proteins represented as networks, residues as nodes and interactions as edges). [14, 15]

In recent years, a number of computational methods have been developed to predict hot spots. These methods are classified into two types: (1) energy-based; and (2) structure-based. In

the energy-based methods, functions are developed to calculate a residue's  $\Delta\Delta G$  by simulating residue mutation to alanine. [19, 20, 21, 22, 23] These methods give good qualitative prediction results. However, high computational cost and the difficulty in operation (e.g. data processing) make them unsuitable for easy implementation. A good example of structure-based methods is the one described by Gao and colleagues. [24] In this method, interface residues are covered by a grid box and the contribution by each residue to binding affinity is estimated by rolling different kinds of probes (representing hydrophobic group, hydrogen bonds) over the grids close to the residue. Thus, residues having high energy contribution are predicted as hot spots. This method is subject to complex structural analysis and comparison. Despite these developments, a simple, robust 'hot spot' prediction model is still unavailable. Here, we describe the analysis and the grouping of 296 alanine-scanned interface residues into three types (hot spots, warm and unimportant residues) towards the development of a novel hot spot prediction model.

#### Methodology:

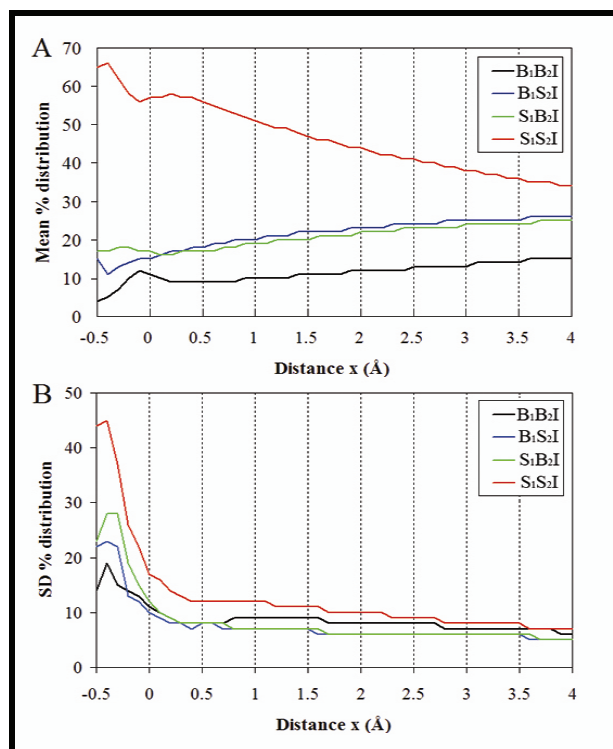
##### Definition of interface residues

ASA (Solvent-accessible surface area) of a residue was calculated using the program NACCESS. [25] A residue with an interface area ( $\Delta\text{ASA}$ )  $> 1 \text{ \AA}^2$  is defined as an interface residue and  $\Delta\text{ASA}$  is the change in ASA of the residue upon protein dimer formation from monomer state.

##### Dataset of alanine-mutated interface residues

A dataset of 296 alanine-mutated interface residues (Supplementary table 1: column 1, 2 and 3) derived from 15 protein-protein complexes (Supplementary table 2) was obtained from ASEdb (Alanine Scanning Energetics database). [26] These residues have  $\Delta\Delta G$  in the range  $-0.9 - 10 \text{ kcal}\cdot\text{mol}^{-1}$ . The dataset was classified into three groups: hot spots ( $\Delta\Delta G \geq 1.5 \text{ Kcal}\cdot\text{mol}^{-1}$ ), warm residues ( $0.5 - 1.5 \text{ Kcal}\cdot\text{mol}^{-1}$ ) and unimportant residues ( $< 0.5 \text{ Kcal}\cdot\text{mol}^{-1}$ ), as

described by Gao *et al.*, [24]



**Figure 1:** The distributions (A: Mean distributions; B: standard deviation) for four categories of inter-atomic interactions ( $B_1B_2I/B_1S_2I/S_1B_2I/S_1S_2I$ ) at the protein-protein interfaces in a non-redundant dataset described elsewhere. [5] B = backbone, S = side-chain, subscript '1' refers to large monomer (e.g. enzymes, antibodies), and subscript '2' refers to small monomer (e.g. inhibitors, antigens). By definition, two atoms from different monomers were considered to interact if the distance between their centers is less than the sum of their van der Waals (vdW) radii plus  $x$  (Å). The value of  $x$  is varied from -0.5 to 4 (Å) at increments of 0.1 (Å). The vdW radius is taken from. [37]

### Definition of inter-atomic sidechain-sidechain interactions

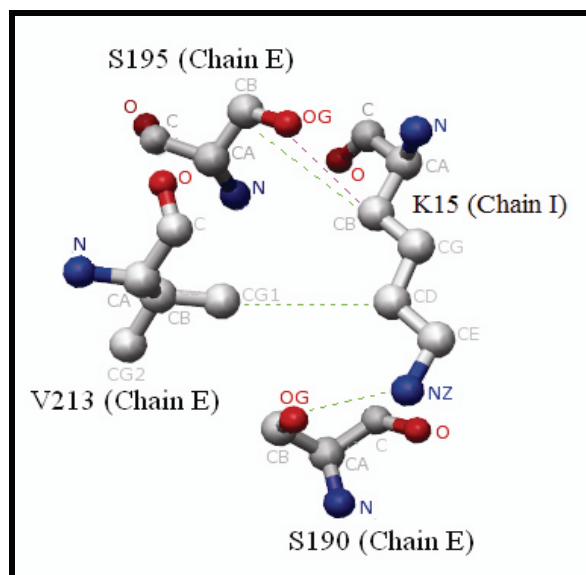
Protein-protein complexation is determined by inter-atomic interactions between monomers. Hence, we investigated the three groups of residues (hot spots, warm and unimportant residues) in terms of their contribution to the inter-atomic interactions. The inter-atomic interactions are composed of four categories, namely  $S_1S_2I$ ,  $S_1B_2I$ ,  $B_1S_2I$  and  $B_1B_2I$  (S: sidechain atom, B: backbone atom, subscript 1 and 2 refer to different monomers). The prevalence of these four inter-atomic interactions at the interface of protein-protein complexes (70 non-redundant complexes [5]) was examined by calculating their means and standard deviations at varying inter-atomic distances (Figure 1).  $S_1S_2I$  dominates at the interface and hence we exclusively selected  $S_1S_2I$  for studying hot spots, warm and unimportant residues. By definition, two sidechain atoms from different monomers were considered to interact if the distance between their centers is less than the sum of their van der Waals (vdW) radii plus a cutoff distance of 0.5 Å, at which cutoff the mean of  $S_1S_2I$  is maximum and the standard deviation is minimum (Figure 1).

### Classification of inter-atomic sidechain-sidechain contacts

We classified inter-atomic sidechain-sidechain contacts into two groups (energetically favorable and unfavorable contacts) using the scheme described by Sobolev and colleagues [27] (Table 1).

### Definition of $NA$ , $NA_I-NA_{II}$ and $NC_I-NC_{II}$

We investigated each residue in the dataset using (1)  $NA$ , (2)  $NA_I-NA_{II}$  and (3)  $NC_I-NC_{II}$  (Supplementary table 1), illustrated in Figure 2.  $NA$  is the number of atoms of a residue participating in sidechain-sidechain contacts.  $NA_I-NA_{II}$  is the difference between the number of atoms in favorable contacts ( $NA_I$ ) and unfavorable contacts ( $NA_{II}$ ). It was employed to explore energetic contribution for a residue to protein-protein interface in terms of atoms.  $NC_I-NC_{II}$  is the difference between favorable contacts ( $NC_I$ ) and unfavorable contacts ( $NC_{II}$ ). It was used to explore energetic contribution for a residue to protein-protein interface in terms of inter-atomic contacts.

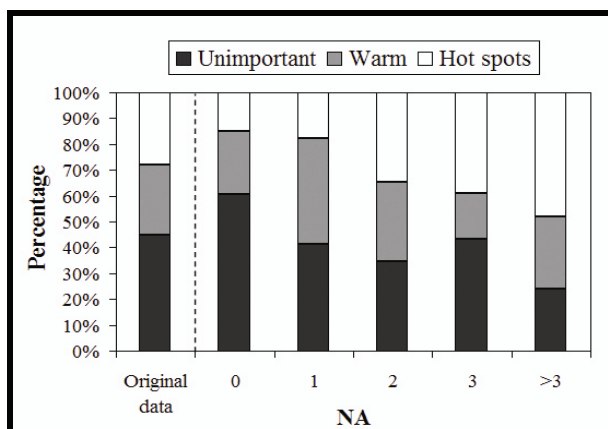


**Figure 2:** Illustration of  $NA$ ,  $NA_I$ ,  $NA_{II}$ ,  $NC_I$  and  $NC_{II}$ . The interaction of K15 (PDB ID: BPTI, Chain I) to S190, S195 and V213 (trypsin, Chain E) is shown (PDB ID: 2PTC). K15 has three interacting side-chain atoms (CB, CD and NZ) and therefore the  $NA$  value is 3. Therein, the three atoms are all involved in favorable contacts (green line) and only CB participates in unfavorable contacts (red line). Thus, the  $NA_I$  value is 3 and  $NA_{II}$  is 1. In addition, K15 has three favorable contacts and one unfavorable contact; hence  $NC_I$  is 3 and  $NC_{II}$  is 1. Carbon atom: white; oxygen atom: red; Nitrogen atom: blue

### Results:

Our goal is to investigate the characteristics of hot spots by comparing them with other interface residues using inter-atomic interactions. We collected 296 alanine-scanning interface residues consisting of 83 hot spots, 80 warm residues and 133 unimportant residues. At the interfaces of subunit interactions,  $S_1S_2I$  (side chain – side chain interaction) dominates and thus,  $S_1S_2I$  was subsequently used in this study. It should be noted that GLY (lacking side

chains) was disregarded in this analysis. However, the current dataset contains only two Gly residues and neither of them is a hot spot. Thus, the elimination of Gly did not significantly effect the analysis. For each residue in the dataset, we calculated the number of atoms (NA) participating in  $S_1S_2I$ , the number of atoms involved in favorable contacts ( $NA_f$ ) and unfavorable contacts ( $NA_{il}$ ). The number of favorable contacts ( $NC_f$ ) and unfavorable contacts ( $NC_{il}$ ) were further calculated. We used these values to calculate  $NA$ ,  $NA_f - NA_{il}$  and  $NC_f - NC_{il}$  for each residue to compare the difference between hot spots, warm and unimportant residues (Supplementary Table 1: column 5, 6 and 7).



**Figure 3:** Percentage distribution of hot spots, warm and unimportant residues in 296 interface residues obtained from ASEdb (alanine scanning energetics database) [26], based on the value NA (the number of atoms for a residue involved in side-chain-side-chain interactions across protein-protein interface). The first column shows the percentage of the three types in the 296 residues. The number of residues is 114 for  $NA=0$ , 34 for  $NA=1$ , 52 for  $NA=2$ , 46 for  $NA=3$  and 50 for  $NA>3$ . White: hot spots; gray: warm residues; black: unimportant residues

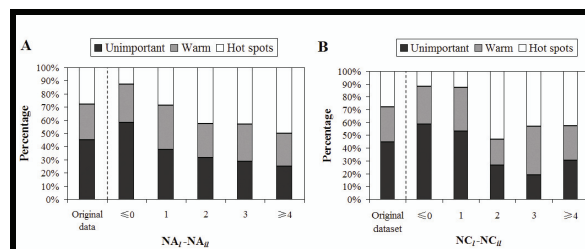
### NA

Figure 3 shows percentage distributions of the three types of interface residues (hot spots, warm and unimportant residues) based on the value of NA. The percentage of hot spots increases from 15% to 50% with NA, while that of unimportant residues decreases from 60% to 23%. Interestingly, the percentage of warm residues does not significantly change with NA. This suggests that  $S_1S_2I$  interactions are prominent among hot spots. When  $NA=1$ , the percentage of warm residues (41%) are larger than that in the original dataset (27%) and when  $NA>1$  hot spots (>33%) are higher than that in the original dataset (28%). It should be noted that nearly 40% of the residues in the dataset do not participate in inter-atomic sidechain-sidechain contacts ( $NA=0$ ). Hence, these residues can not be identified as hot spots, warm and unimportant residues using NA,  $NA_f - NA_{il}$  and  $NC_f - NC_{il}$  values.

### $NA_f - NA_{il}$

Figure 4A shows percentage distributions of hot spots, warm and unimportant residues based on  $NA_f - NA_{il}$ . The percentages of unimportant residues decrease with the increase in  $NA_f - NA_{il}$ , and that of hot spots increases. The percentage of warm residues is not significantly affected by  $NA_f - NA_{il}$ . We

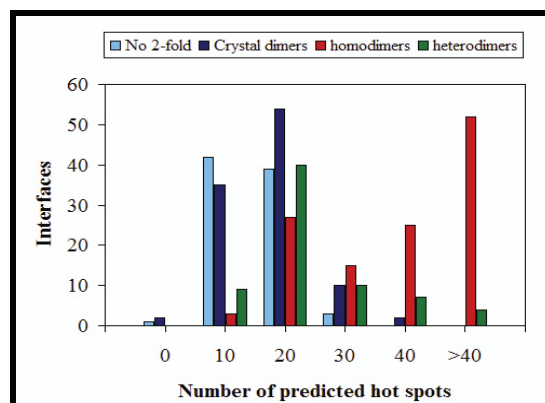
also show that when  $NA_f - NA_{il} > 1$ , the percentage of hot spots is larger than the fraction of hot spots in the original dataset (28%).



**Figure 4:** Percentage distribution of the three residue types in 182 interface residues with  $NA>0$  (Unimportant residues: 64; warm residues: 52; hot spots: 66) based on the value of (A)  $NA_f - NA_{il}$  (the difference between numbers of sidechain atoms for a residue involved in favorable and unfavorable contacts). The number of residues is for 24 for  $NA_f - NA_{il} < 1$ , 45 for  $NA_f - NA_{il} = 1$ , 54 for  $NA_f - NA_{il} = 2$ , 35 for  $NA_f - NA_{il} = 3$  and 24 for  $NA_f - NA_{il} > 4$  (B)  $NC_f - NC_{il}$  (the difference between numbers of favorable and unfavorable contacts). The number of residues is 17 for  $NC_f - NC_{il} < 1$ , 32 for  $NC_f - NC_{il} = 1$ , 30 for  $NC_f - NC_{il} = 2$ , 21 for  $NC_f - NC_{il} = 3$ , 82 for  $NC_f - NC_{il} > 3$ . The first column in each graph shows the percentages of the three types in 296 interface residues. White: hot spots; gray: warm residues; black: unimportant residues

### $NC_f - NC_{il}$

Figure 4B shows percentage distribution of the three types of interface residues types based on  $NC_f - NC_{il}$ . The percentage of unimportant residues decreases with the increase in  $NC_f - NC_{il}$ , and hot spots increases. The percentage of warm residues does not significantly change with  $NC_f - NC_{il}$ . It was also found that the percentage of hot spots is high when  $NC_f - NC_{il} \geq 2$ , in comparison to the fraction (28%) of hot spots in the original dataset.



**Figure 5:** Distribution of hot spots calculated using our 'hot spot' prediction method for non-specific complexes (crystal-packing artifacts) and specific complexes (homodimers and heterodimers). These complexes are derived from the dataset of Bahadur *et al.*, [36] Cyan, large crystal packing interfaces with no 2-fold symmetry; blue, crystal dimers; red, homodimers; green, protein-protein complexes

### Discussion:

#### A 'hot spot' prediction approach

Results show that the fraction of hot spots increases and unimportant residues decreases with increase in NA,  $NA_{I-NA_{II}}$  and  $NC_{I-NC_{II}}$ . However, the fraction of warm residues is not significantly affected by these three parameters. Thus, hot spots are preferentially involved in strong and energetically favorable sidechain-sidechain interactions, unimportant residues tend to participate in weak and energetically unfavorable sidechain-sidechain interactions. Here, we used NA,  $NA_{I-NA_{II}}$  and  $NC_{I-NC_{II}}$  to develop a method to identify hot spots using interface residues in structural complexes. We classified the residues in our dataset using a combination of three parameters. This is based on the observation that hot spots are prevailing in residues with  $NA > 1$ ,  $NA_{I-NA_{II}} > 1$  or  $NC_{I-NC_{II}} > 1$ , and unimportant residues are predominant in those with  $NA = 0$ ,  $NA_{I-NA_{II}} \leq 1$  or  $NC_{I-NC_{II}} \leq 1$  (Figure 3 and 4). Table 2 shows that the percentages of unimportant residues when (i)  $NA = 0$ , (ii)  $NA = 1$  &&  $NA_{I-NA_{II}} \leq 1$  &&  $NC_{I-NC_{II}} \leq 1$ , and (vi)  $NA > 0$  &&  $NA_{I-NA_{II}} \leq 1$  &&  $NC_{I-NC_{II}} \leq 1$  are larger than that in original dataset; and hot spots in (iii)  $NA = 1$  &&  $NA_{I-NA_{II}} \leq 1$  &&  $NC_{I-NC_{II}} \geq 2$ , (vii)  $NA > 1$  &&  $NA_{I-NA_{II}} \leq 1$  &&  $NC_{I-NC_{II}} \geq 2$ , and (ix)  $NA > 1$  &&  $NA_{I-NA_{II}} \geq 2$  &&  $NC_{I-NC_{II}} \geq 2$  are higher than original dataset. Thus, the residues with  $NC_{I-NC_{II}} \geq 2$  could be predicted as hot spots and those with  $NC_{I-NC_{II}} \leq 1$  as unimportant residues. Therefore, these observations find application in the development of an expert system for the identification of hot spots from structural complexes.

Atom class*		Favorable (+) or unfavorable (-) contact							
		I	II	III	IV	V	VI	VII	VIII
I	Hydrophilic	+	+	+	-	+	+	+	+
II	Acceptor	+	-	+	-	+	+	+	-
III	Donor	+	+	-	-	+	+	-	+
IV	Hydrophobic	-	-	-	+	+	+	+	+
V	Aromatic	+	+	+	+	+	+	+	+
VI	Neutral	+	+	+	+	+	+	+	+
VII	Neutral-donor	+	+	-	+	+	+	-	+
VIII	Neutral-acceptor	+	-	+	+	+	+	+	-
or									

**Table 1:** Legitimacy of contacts between side-chain atoms in different classes

\*I: Hydrophilic = nitrogen or oxygen atoms that can donate and accept hydrogen bonds. II: Acceptor = nitrogen or oxygen atoms that can only accept a hydrogen bond. III: Donor = nitrogen that can only donate a hydrogen bond. IV: Hydrophobic = carbon atoms that are not in aromatic rings and do not have a covalent bond to a hydrophilic atom. V: Aromatic = carbon atoms in aromatic rings. VI: Neutral = carbon atoms that have a covalent bond to at least one atom of class I or two or more atoms from class II or III; nitrogen atoms if it has covalent bonds with 3 carbons; sulfur atoms in all cases. VII: Neutral-donor = carbon atoms that has a covalent bond with only one atom of class III. VIII: Neutral-acceptor = carbon atoms that has covalent bond with only one atom of class II. The classification is derived from. [27]

#### Comparison with other 'hot spot' predication methods

We evaluated our 'hot spot' prediction method by comparing them with three other methods: (1) PP\_SITE [24], (2) alanine scanning method developed by Kortemme and coworkers [20]

and (3) FOLDEF. [22] The PP\_SITE method is structure-based, while the other two are energy-based. We assessed the performance of the four methods in distinguishing hot spots and unimportant residues. The PP\_SITE classified interface residues into three types (hot spots, warm and unimportant residues) and its predicted warm residues include 43% of experimental hot spots. [24] In Alanine Scanning and FOLDEF methods, we considered interface residues with calculated  $\Delta\Delta G \geq 1$  Kcal·mol<sup>-1</sup> as predicted hot spots and other residues as predicted unimportant residues.

The four methods were evaluated using our dataset of 296 interface residues. The FOLDEF and our method identified all the 296 residues, while the PP\_SITE method identified 226 residues and alanine scanning method identified 261 residues (See supplementary table 1). Then, we retained the identified residues which belong to experimental hot spots and unimportant residues (FOLDEF and our method: 215 residues; PP\_SITE: 160 residues; Alanine scanning: 187 residues). Finally, for each method, we calculated sensitivity (SN), specificity (SP), positive predictive value (PPV), negative predictive value (NPV) and average successful rate ((TP+TN)/(TP+TN+FN+FP)) for hot spot prediction (Table 3).

Our method and FOLDEF showed high average successful rate (71% - 72%), compared to the other two methods (PP\_SITE: 66%; Alanine Scanning: 68%). Thus, the FOLDEF and our method can effectively distinguish between hot spots and unimportant residues. Our method efficiently identified hot spots (SN = 72%; SP = 72%), while the FOLDEF efficiently identified unimportant residues (SN = 45%; SP = 88%). In addition, the PP\_SITE correctly identified most host spots (SN = 90%) in these methods. However, it could not effectively differentiate unimportant residues from hot spots (SP = 37%). It agreed with the conclusions drawn by Gao *et al.* that the PP\_SITE over-estimated unimportant residues. From these analyses, we can see that our method has remarkable hot spot prediction accuracy relative to the prevailing prediction approaches.

#### Misidentified hot spots

Out of the 83 hot spots in our dataset, 23 were not predicted, 17 of which do not have sidechain-sidechain interactions ( $NA = 0$ ) and the remaining five do not make significant energetic contribution to sidechain-sidechain interactions ( $NC_{I-NC_{II}} = 1$ ). It seems that energetic contribution of these hot spots to protein-protein interaction could not be reflected by their participation in inter-monomeric sidechain-sidechain interactions. In order to understand how the 23 misidentified hot spots contribute to protein-protein interaction, they were studied in detail and several reasons were found. (1) Some of them interact with interfacial water molecules to enhance the stability of protein-protein interaction. For instance, the residue D51 in the protein Im9 (PDB: 1BXI) hydrogen bonds two interfacial water molecules buried in cavities. [28] (2) Some neighbor with hot spots with  $NC_{I-NC_{II}} \geq 2$ . The mutations to alanine may influence their neighboring hot spot's conformation which then reduces protein-protein interaction. In human growth hormone-hGH receptor complex (PDB: 3hhr), misidentified



hot spots I103 ( $\Delta\Delta G = 1.8 \text{ Kcal}\cdot\text{mol}^{-1}$ ) and I105 ( $\Delta\Delta G = 2.0 \text{ Kcal}\cdot\text{mol}^{-1}$ ) neighbor hot spot W104 ( $\Delta\Delta G = 4.5 \text{ Kcal}\cdot\text{mol}^{-1}$ ). (3) Some have role in stabilizing monomer structure so that their mutations to alanine disrupt monomer conformation which weakens protein-protein interactions, such as the residue D58 in Tissue Factor (PDB: 1DAN). [29] (4) Some contribute to protein-protein interaction by participating in backbone-backbone or backbone-sidechain interactions. The

residue K15 in protein Basic Pancreatic Trypsin Inhibitor (PDB: 1CBW) are involved in three backbone-backbone hydrogen bonds. [30] Similarly, the residues N23 and Q120 in Staphylococcal enterotoxin C3 (PDB: 1JCK) form hydrogen bonds with backbone atoms of interacting monomer T cell antigen receptor V $\beta$ . [31]

Groups	Interface residues			Total (296)
	Hot-spot 83 (28%)	Warm 80 (27%)	Unimportant 133 (45%)	
(i) NA=0	17 (15%)	28 (25%)	69 (60%)	114
NA=1				
(ii) NA <sub>T</sub> -NA <sub>ii</sub> ≤1 & NC <sub>T</sub> -NC <sub>ii</sub> ≤1	2 (8%)	10 (40%)	13 (53%)	25
(iii) NA <sub>T</sub> -NA <sub>ii</sub> ≤1 & NC <sub>T</sub> -NC <sub>ii</sub> ≥2	4 (44%)	4 (44%)	1 (11%)	9
(iv) NA <sub>T</sub> -NA <sub>ii</sub> ≥2 & NC <sub>T</sub> -NC <sub>ii</sub> ≤1	0	0	0	0
(v) NA <sub>T</sub> -NA <sub>ii</sub> ≥2 & NC <sub>T</sub> -NC <sub>ii</sub> ≥2	0	0	0	0
NA>1				
(vi) NA <sub>T</sub> -NA <sub>ii</sub> ≤1 & NC <sub>T</sub> -NC <sub>ii</sub> ≤1	4 (17%)	6 (25%)	14 (58%)	24
(vii) NA <sub>T</sub> -NA <sub>ii</sub> ≤1 & NC <sub>T</sub> -NC <sub>ii</sub> ≥2	6 (55%)	2 (18%)	3 (27%)	11
(viii) NA <sub>T</sub> -NA <sub>ii</sub> ≥2 & NC <sub>T</sub> -NC <sub>ii</sub> ≤1	0	0	0	0
(ix) NA <sub>T</sub> -NA <sub>ii</sub> ≥2 & NC <sub>T</sub> -NC <sub>ii</sub> ≥2	50 (44%)	30 (27%)	33 (29%)	113

**Table 2:** Classification of the residues in the datasets using the three parameters (NA, NA<sub>T</sub>-NA<sub>ii</sub> and NC<sub>T</sub>-NC<sub>ii</sub>)

	Our approach	PP_SITE [24]	Alanine Scanning [20]	FOLDEF [22]
SN	72%	90%	60%	45%
SP	72%	47%	74%	88%
PPV	62%	57%	64%	70%
NPV	81%	86%	71%	72%
Averaged successful rate	72%	66%	68%	71%
% of warm residues predicted as hot spots	45%	60%	42%	21%

**Table 3:** Evaluation of 'hot spot' prediction approaches

The four prediction methods were assessed by comparing their performance on the differentiation between hot spots and unimportant residues. Warm residues were disregarded. SN=sensitivity; SP=specificity; PPV= positive predictive value; NPV= negative predictive value; average successful rate = ((TP+TN) / (TP+TN+FN+FP)). Both predicted warm residues and hot spots by the method PP\_SITE were regarded as predicted hot spots here and the evaluation is based on the PP\_SITE prediction result with surface punishment. [24] For the alanine scanning method and the FOLDEF method, we considered interface residues with calculated  $\Delta\Delta G \geq 1 \text{ Kcal}\cdot\text{mol}^{-1}$  as predicted hot spots and other residues as predicted unimportant residues.

### Distinction between specific and non-specific complexes

Assessing the oligomeric state of a protein from its X-ray structure is not always straightforward and protein subunit interfaces often coexist with 6 to 12 packing interfaces. [32, 33] The distinction between oligomers (specific complexes) and crystal-packing artifacts (non-specific complexes) is often made on the basis of interface area and specific interface area is generally larger. [3, 34, 35] Recently, Bahadur *et al.* observed that three independent parameters (non-polar interface area, fraction of fully buried atoms and residue propensity score at interface) could distinguish between homo-dimers and non-specific complexes and these are indistinguishable based on interface area. [36] Here, we used our 'hot spot' prediction method to distinguish between specific and non-specific complexes, using the dataset of Bahadur *et al.* which contains 188 large crystal-packing artifacts, 122 homo-dimers and 70 hetero-dimers. Figure 5 show that the low abundance of hot spots distinguishes the crystal-packing interfaces from homo-dimeric interfaces. Using the number (23) of hot

spots as a cutoff, 179 out of 188 non-specific interfaces and 88 out of 122 homo-dimeric interfaces were identified. In other words, 86% of the proteins are correctly classified as monomers and homo-dimers using hot spots as a criterion. The hot spot cutoff was selected manually in this study and with larger data sets, the cutoff has to be refined to optimized, for the distinction between homo-dimers and monomers. We also calculated the correlations between the number of hot spots and the three parameters observed by Bahadur *et al.* and found a weak correlation (correlation coefficient  $R^2 < 0.17$ ). Thus, the 'hot spot' prediction method could be applied along with these three parameters for homo-dimer identification. However, the prediction method could not efficiently distinguish between hetero-dimers and non-specific complexes. This may be due to the binding mechanism of hetero-dimers, which assemble from preformed protein components. In the free components, the surface patches that form the interface are in contact with the solvent and their physical/chemical properties are not significantly different from the remainder of protein

surface.

### Electronic supplementary material

The dataset of 296 alanine-mutated interface residues in this work is available online.

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**Supplementary Table 1: Analysis of 296 mutated interface residues**

PDB ID	Residue	$\Delta\Delta G$ (Kcal/Mol)	$\Delta ASA$ (Å <sup>2</sup> )	NA	$NA_I-NA_{II}$	$NC_I-NC_{II}$	Alanine scanning (Kcal/Mol)	FOLDEF (Kcal/Mol)	PP_SITE
1a4yA	R5	2.3	100.1	3	-1	-1	2.54	2.09	H
1a4yA	H8	0.9	24.1	2	2	2	0.84	-0.37	U
1a4yA	Q12	0.3	22.8	2	2	3	0.66	0.4	U
1a4yA	H13	-0.3	4.7	0	0	0	0.02	0.19	U
1a4yA	R31	0.2	119.9	3	2	8	2.76	-0.83	W
1a4yA	R32	0.9	62.2	3	3	9	0.18	0.06	U
1a4yA	N68	0.2	11.1	0	0	0	0.52	0.05	U
1a4yA	H84	0.2	64	3	3	5	1.05	0.25	W
1a4yA	W89	0.2	84.8	5	5	13	2.71	0.88	H
1a4yA	E108	-0.3	43.9	0	0	0	1.58	-0.15	H
1a4yA	H114	0.65	62.6	5	3	3	1.69	0.49	W
1a4yB	W261	0.1	32	0	0	0	1.06	0.6	H
1a4yB	W263	1.2	66	2	2	4	2.27	2.2	H
1a4yB	W318	1.5	42.5	7	7	12	2.2	2.7	H
1a4yB	K320	-0.3	39.5	1	1	1	-0.21	-0.75	U
1a4yB	E344	0.2	16	3	3	6	1.37	-0.44	U
1a4yB	W375	1	56.6	7	5	12	2.83	2.51	H
1a4yB	E401	0.9	19.6	1	1	1	0.02	-0.2	W
1a4yB	Y434	3.3	98.5	6	6	8	2.95	1.93	H
1a4yB	D435	3.5	80.3	3	1	6	0.57	3.2	H
1a4yB	Y437	0.8	143.4	5	4	7	2.98	1.07	H
1a4yB	R457	-0.2	2.8	0	0	0	-0.04	0.32	-
1a4yB	I459	0.7	23.9	1	1	1	0.62	0.41	W
1ahwA	Y156	>4	65.9	3	2	6	4.6	2.00	W
1ahwA	T167	0	41.5	0	0	0	-0.03	0.05	U
1ahwA	T170	1	11.3	0	0	0	-0.06	0.45	U
1ahwA	D178	-0.5	14.3	0	0	0	-0.09	0.15	U
1ahwA	T197	1.3	8.9	0	0	0	-0.02	0.01	U
1ahwA	V198	-0.3	7.2	0	0	0	-0.01	-0.03	-
1ahwA	N199	1.1	0.7	0	0	0	-0.01	0.03	-
1brsA	K27	5.4	54.6	3	3	5	1.88	2.58	H
1brsA	N58	3.1	2.6	0	0	0	-0.03	0.73	U
1brsA	R59	5.2	144.2	7	1	10	3.01	0.83	H
1brsA	E60	-0.2	65.4	3	0	-2	1.4	0.43	W
1brsA	E73	2.8	9.5	0	0	0	-0.2	-0.59	W
1brsA	R87	5.5	5.2	1	1	2	4.44	-0.12	W
1brsA	H102	6	95.3	6	4	10	5.08	1.81	W
1brsB	Y29	3.4	86.2	4	4	13	3.13	1.86	H
1brsB	D35	4.5	111	4	1	7	1.42	0.54	H
1brsB	D39	7.7	80.9	4	3	13	9.4	3.07	H
1brsB	T42	1.8	16	2	2	5	1.66	2.01	W
1brsB	E76	1.3	21.7	3	3	5	1.54	0.12	U
1bxiA	C23	0.92	10.6	0	0	0	-0.08	-0.10	U
1bxiA	N24	0.14	6.4	0	0	0	0.00	-0.05	U
1bxiA	T27	0.73	5.3	1	1	3	0.59	0.90	U
1bxiA	S28	0.17	0	0	0	0	0.00	0.04	U
1bxiA	S29	0.96	6.6	0	0	0	0.35	-0.04	U
1bxiA	E30	1.41	100.5	5	3	11	2.97	0.94	H

1bxiA	L33	3.42	33.4	2	1	4	1.02	1.07	W
1bxiA	V34	2.58	61.5	1	1	2	0.98	0.91	H
1bxiA	V37	1.66	16.2	3	3	4	0.5	0.44	W
1bxiA	T38	0.9	19.2	1	0	0	1.35	1.00	W
1bxiA	E41	2.08	32.8	2	2	3	-0.08	0.58	H
1bxiA	S48	0.01	2.1	0	0	0	-0.01	-0.05	U
1bxiA	G49	1.49	8.6	0	0	0	-	-0.43	U
1bxiA	S50	2.19	33	1	1	2	5.4	1.39	W
1bxiA	D51	5.92	36.6	0	0	0	0.82	1.07	W
1bxiA	Y55	4.63	90.6	4	4	6	2.8	3.31	H
1bxiA	P56	1.24	5.4	0	0	0	-	0.14	U
1cbwA	T11	0.2	37.9	0	0	0	-0.18	0.06	U
1cbwA	P13	-0.1	46.3	1	1	1	-	0.75	W
1cbwA	K15	2	147.6	2	-1	-1	-0.03	-0.74	H
1cbwA	R17	0.5	151.1	6	3	9	1.28	1.22	H
1cbwA	I19	0.1	38.1	0	0	0	0.68	0.40	W
1cbwA	V34	0	14.1	0	0	0	0.3	0.40	W
1cbwA	R39	0.2	64.8	1	1	1	1.61	1.17	W
1danA	L39	0	103.2	3	3	7	1.24	1.10	-
1danA	I42	0	3.5	0	0	0	0.51	0.23	-
1danA	K62	0	72	2	0	2	-0.27	-0.45	-
1danA	Q64	0.8	60.5	1	1	1	2.98	0.30	-
1danA	I69	1.9	74.7	2	2	4	1.57	1.66	-
1danA	F71	1.2	100.2	5	5	13	3.07	2.58	-
1danA	E77	0	60.3	2	0	0	0.36	-1.19	-
1danA	R79	1.2	110.4	4	2	8	2.45	0.34	-
1danA	Q88	0	29.4	3	3	5	0.35	-0.05	-
1danA	V92	0	65.6	1	1	4	1.05	0.97	-
1danA	N93	0	29.3	3	3	6	0.12	0.07	-
1danA	E94	0	38.3	0	0	0	-	-0.13	-
1danA	R271	0	1.3	0	0	0	-	0.03	-
1danA	F275	0	71.2	7	7	9	1.66	1.32	-
1danA	R277	0.51	120.6	1	0	3	3.09	0.95	-
1danA	F278	0	25.8	0	0	0	0.53	0.27	-
1danA	R304	0.65	2.9	0	0	0	-	0.88	-
1danA	M306	0.5	110.8	4	0	11	1.03	1.08	-
1danA	T307	0	6	0	0	0	0.27	-1.30	-
1danA	Q308	0	72.8	3	2	5	2.6	0.66	-
1danA	D309	0.41	42.5	3	3	4	1.34	1.52	-
1danA	Q312	0	5	0	0	0	-	-0.14	-
1danA	E325	0	3.9	0	0	0	-	0.04	-
1danA	R379	0.51	30.9	3	2	5	1.33	-0.93	-
1danB	K15	-0.4	3.3	0	0	0	-0.02	-0.08	-
1danB	T17	0.1	25.7	2	2	2	0.12	0.22	U
1danB	N18	0.2	11.3	0	0	0	0.04	0.19	U
1danB	K20	2.6	75.9	2	2	2	1.5	0.16	H
1danB	I22	0.7	27.7	2	0	0	0.65	0.64	W
1danB	E24	0.7	14.7	2	2	2	0.64	-0.44	U
1danB	Q37	0.55	27.3	1	1	6	0.75	1.37	W
1danB	K41	0.35	3.2	0	0	0	-0.04	-0.07	U
1danB	S42	-0.1	18.1	0	0	0	-0.05	-0.08	U



1danB	D44	0.7	51.2	0	0	0	0.9	0.89	W
1danB	W45	1.6	49.1	4	2	2	1.05	0.97	H
1danB	K46	0.25	39.1	0	0	0	0.10	0.22	W
1danB	S47	0.05	22.9	0	0	0	-0.05	-0.19	W
1danB	K48	0.4	22.8	2	2	2	0.43	0.04	W
1danB	F50	0.4	101.2	7	7	17	2.61	3.09	H
1danB	Y51	-0.1	34.1	1	1	1	-	0.42	H
1danB	D58	2.18	43.2	0	0	0	1.09	1.13	W
1danB	D61	0.24	50.3	1	0	1	0.01	0.27	W
1danB	E62	0	1.1	0	0	0	-	0.32	-
1danB	L72	-0.06	0.2	0	0	0	-	-0.12	-
1danB	F76	1.2	16	2	2	3	0.61	1.23	W
1danB	Y78	0.7	0.3	0	0	0	-	-0.01	W
1danB	P92	-0.2	7	0	0	0	-	-0.33	U
1danB	Q110	1.4	58.5	2	1	1	1.69	0.02	U
1danB	E128	0.1	14.5	0	0	0	-0.11	-0.13	U
1danB	R131	0	44.4	2	2	6	0.29	0.37	U
1danB	L133	0	41.6	2	2	5	1.62	1.70	H
1danB	R135	0.55	68.8	0	0	0	0.94	0.23	W
1danB	N138	0	15.8	0	0	0	-	0.02	-
1danB	F140	1.5	13.3	3	3	3	1.54	0.74	H
1danB	S163	0	24.6	2	2	2	0.42	-0.15	U
1danB	T203	0.1	25.6	3	1	1	0.22	0.10	U
1danB	V207	-0.2	55	2	2	3	1.12	1.40	H
1danB	E208	0	25.8	1	1	1	0.28	0.15	U
1dfjB	E206	1	24.1	1	1	2	0.77	-0.97	W
1dfjB	W261	1.3	39.2	2	2	2	1.58	1.39	H
1dfjB	W263	2.2	43.5	3	3	3	3.09	2.53	H
1dfjB	E287	1.3	23.9	1	1	1	-0.1	-0.08	U
1dfjB	S289	0.8	2.2	0	0	0	0	-0.02	U
1dfjB	W318	1	54.7	2	2	3	1.27	1.24	H
1dfjB	K320	1.3	42.2	1	1	1	0.38	-0.07	W
1dfjB	E344	1.6	5.1	0	0	0	-	0.30	-
1dfjB	E401	1.3	26.3	2	2	6	0.42	0.21	W
1dfjB	Y434	5.9	107.7	7	7	14	2.74	4.23	H
1dfjB	D435	3.6	62.7	2	2	2	0.16	1.03	W
1dfjB	Y437	2.6	158.3	7	5	9	3.62	2.62	H
1dfjB	R457	0.8	28.6	2	2	3	0.21	-0.11	U
1dfjB	I459	0.3	35.3	1	1	1	0.5	0.52	W
1dvfA	T30	0.9	18	0	0	0	0.01	0.18	U
1dvfA	Y32	1.8	22.1	0	0	0	0.11	0.19	W
1dvfA	W52	4.2	65.5	7	5	6	3.26	3.52	H
1dvfA	D54	4.3	65.2	3	3	6	0.72	0.86	W
1dvfA	N56	1.2	53.2	3	1	3	1.07	-0.68	U
1dvfA	D58	1.6	14.1	3	3	3	0.71	0.28	W
1dvfA	E98	4.2	7.4	3	2	5	0.05	1.76	U
1dvfA	R99	1.9	60	0	0	0	0.22	0.94	U
1dvfA	D100	2.8	47	3	2	2	0.43	0.69	W
1dvfA	Y101	4	68	4	2	2	2.13	1.62	H
1dvfA	H1030	1.7	18	1	1	1	0.26	0.20	W
1dvfA	Y1032	2	44.3	2	2	2	0.47	0.24	H

ldvfA	Y1049	1.7	23.1	3	3	3	0.87	0.43	W
ldvfA	Y1050	0.7	53	3	3	3	0.55	-0.11	H
ldvfA	W1092	0.3	55.8	4	4	6	1.56	0.68	H
ldvfA	S1093	1.2	11.4	0	0	0	-0.02	0.01	U
ldvfB	H33	1.9	27	2	2	3	0.79	0.46	W
ldvfB	D52	1.7	15.8	0	0	0	-0.18	-0.29	W
ldvfB	N55	1.9	46.2	1	1	3	2.13	-0.53	W
ldvfB	I100	2.7	49.5	1	1	1	1.07	1.29	H
ldvfB	Y101	4.7	165.7	4	3	11	4.77	5.42	W
ldvfB	Q103	1.6	90.6	5	2	9	1.42	1.91	W
ldvfB	R105	4.1	94.4	5	3	6	1.5	1.44	H
lfc2A	N28	0.6	46	3	0	0	0.29	0.92	-
lfc2A	I31	2.2	25.5	2	1	1	0.81	0.32	-
lfc2A	K35	1.2	59.7	2	0	0	0.18	-0.13	-
lgl1A	S23	0.29	4.2	0	0	0	0.27	-0.15	W
lgl1A	Q25	0.03	20.1	1	0	1	0.42	2.82	W
lgl1A	H27	0.28	38.6	5	1	1	0.8	0.44	W
lgl1A	K29	0.59	24.9	2	2	5	2.46	-0.07	W
lgl1A	N32	0.18	3	0	0	0	0	0.02	U
lgl1A	Q33	0.1	17.3	0	0	0	0.03	-0.05	W
lgl1A	K35	0.32	83.9	4	-1	-1	0.55	-1.19	U
lgl1A	Q40	-0.41	79	0	0	0	1.66	4.17	W
lgl1A	S42	0	67.5	0	0	0	0.03	-1.06	W
lgl1A	L44	1.04	7.8	0	0	0	0.07	0.43	U
lgl1A	T45	-0.15	23.7	1	1	1	0.32	0.65	U
lgl1A	N52	0.7	23.3	0	0	0	1.01	-0.04	U
lgl1A	R59	1.16	47.2	4	4	9	1.02	0.45	W
lgl1A	S60	-0.09	39.8	0	0	0	0.14	-0.36	U
lgl1A	D63	-0.32	37.2	0	0	0	-0.05	0.49	W
lgl1A	Q64	0.44	29.6	3	1	1	1.02	-0.47	W
lgl1A	E85	1.31	1.8	0	0	0	2.1	1.02	U
ljckA	T20	1.4	59.7	0	0	0	1.26	0.33	W
ljckA	N23	2.5	49	0	0	0	1.96	0.27	W
ljckA	Y26	1.7	38.7	0	0	0	0.92	0.40	H
ljckA	N60	1.3	42.3	0	0	0	0.82	0.40	W
ljckA	Y90	2.5	31.7	2	2	2	1.0	0.79	H
ljckA	V91	2.1	71.7	2	2	2	1.04	0.93	H
ljckA	G102	0.1	14.1	0	0	0	-	-0.30	U
ljckA	K103	0.4	19.1	0	0	0	-0.26	-1.09	W
ljckA	F176	1.9	44.5	2	2	2	0.63	-0.16	H
ljckA	Q210	2.5	14.1	0	0	0	1.12	0.24	U
lvfbA	Y32	0.5	30.5	1	1	1	0.66	-0.09	W
lvfbA	W52	1.23	53.8	4	4	7	1.75	1.55	H
lvfbA	D54	1.95	32.5	2	1	1	-0.14	0.12	U
lvfbA	R99	0.47	49	0	0	0	0.73	0.51	U
lvfbA	D100	3.1	60.3	3	3	6	3.29	2.22	H
lvfbA	Y101	4	69.1	4	3	4	2.63	3.55	H
lvfbA	H1030	0.8	28.7	0	0	0	0.26	0.21	H
lvfbA	Y1032	1.3	55.4	7	7	7	1.32	1.00	H
lvfbA	Y1049	0.8	22.2	0	0	0	0.49	0.50	U
lvfbA	Y1050	0.4	72.1	4	4	10	1.52	1.28	H

1vfbA	T1053	-0.23	17.7	2	2	2	0.64	0.66	U
1vfbA	W1092	1.71	60.9	4	4	6	2.06	3.01	H
1vfbA	S1093	0.11	35.5	0	0	0	0.00	-0.29	U
1vfbB	D18	0.3	31.9	3	3	6	0.55	0.58	W
1vfbB	N19	0.3	72.3	3	3	6	1.04	0.84	W
1vfbB	Y23	0.4	9.3	0	0	0	0.0	0.15	W
1vfbB	S24	0.8	53.4	2	2	4	0.78	-0.16	W
1vfbB	K116	0.7	52.6	1	1	1	0.85	-0.01	W
1vfbB	T118	0.8	33.9	2	1	1	0.1	0.05	U
1vfbB	D119	1	40.9	3	3	8	-0.2	0.35	W
1vfbB	V120	0.9	13.3	1	1	1	0.22	0.44	U
1vfbB	Q121	2.9	95.1	3	2	10	3.87	2.58	W
1vfbB	I124	1.2	23.4	1	1	1	0.46	0.40	W
1vfbB	R125	1.8	56.5	3	3	5	2.25	0.65	W
2ptcA	K15	10	165.8	3	2	2	4.16	0.33	-
3hfmA	H15	-0.44	12.9	0	0	0	0.09	0.14	U
3hfmA	Y20	4.2	66.6	2	2	2	2.72	1.46	H
3hfmA	R21	0.85	118.4	3	3	15	0.94	-1.32	H
3hfmA	W63	0.31	22.9	2	2	4	0.83	0.75	W
3hfmA	R73	-0.33	67.9	5	3	7	0.62	-0.40	W
3hfmA	L75	0.69	59	0	0	0	1.31	0.89	H
3hfmA	T89	0	15.4	0	0	0	0.17	0.08	U
3hfmA	N93	0.21	44	3	2	8	1.56	2.01	U
3hfmA	K96	6.38	51.5	4	0	8	2.13	-1.36	W
3hfmA	K97	5.5	72.2	2	2	4	1.42	0.32	W
3hfmA	I98	-0.10	2.1	0	0	0	0.10	-0.04	U
3hfmA	S100	0.26	23.2	2	2	4	-0.13	-0.11	U
3hfmA	D101	0.94	84.7	4	3	15	0.32	-2.39	H
3hfmB	Y58	1.7	45.9	6	6	6	1.77	0.55	H
3hfmB	D101	3.75	0.5	0	0	0	-0.04	0.09	-
3hhrA	P2	-0.05	75.9	2	2	3	-	0.06	H
3hhrA	T3	-0.05	0.8	0	0	0	-	-0.01	U
3hhrA	I4	0.41	68.4	0	0	0	-	-0.08	H
3hhrA	R8	0.2	88.5	3	2	7	-	-1.19	H
3hhrA	L9	-0.04	14.7	1	1	1	-	-0.01	U
3hhrA	N12	0.1	59.1	4	3	8	-	-0.26	U
3hhrA	L15	0.15	70.5	3	0	1	-	-0.05	H
3hhrA	R16	0.24	80.8	6	5	18	-	-0.11	W
3hhrA	H18	-0.5	65.2	3	0	-1	1.51	-1.32	H
3hhrA	R19	0.05	44.4	0	0	0	-	0.05	W
3hhrA	H21	0.2	25.7	0	0	0	0.64	0.20	W
3hhrA	Q22	-0.2	14.2	0	0	0	0.06	0.36	U
3hhrA	F25	-0.4	63	0	0	0	1.68	1.08	H
3hhrA	D26	-0.2	0.1	0	0	0	-	0.39	-
3hhrA	Q29	-0.6	2.3	0	0	0	-	0.19	-
3hhrA	Y42	0.2	92.7	2	1	0	2.09	1.34	H
3hhrA	L45	1.2	57.7	2	2	2	1.13	1.36	H
3hhrA	Q46	0.1	90.6	3	0	2	1.2	0.17	U
3hhrA	P48	0.4	21.8	2	-1	-1	-	0.56	U
3hhrA	S51	0.3	9.3	0	0	0	-0.04	0.00	-
3hhrA	E56	0.4	1.2	2	2	3	-0.04	0.21	-

3hhrA	S62	0.2	52	0	0	0	0.18	1.03	U
3hhrA	N63	0.3	38.5	0	0	0	0.22	-0.25	W
3hhrA	R64	1.6	115.2	4	3	13	1.12	1.42	W
3hhrA	E65	-0.5	1.2	0	0	0	-0.05	-0.16	-
3hhrA	Q68	0.6	25.2	0	0	0	0.29	-0.22	U
3hhrA	Y164	0.3	25	1	1	1	0.6	-1.08	U
3hhrA	R167	0.3	31.9	2	1	2	0.54	-0.64	W
3hhrA	K168	-0.2	57.3	2	0	0	1.14	-2.41	U
3hhrA	D171	0.8	51	4	2	2	1.4	3.94	H
3hhrA	K172	2	31.8	2	2	8	0.71	0.77	U
3hhrA	E174	-0.9	23.9	1	1	1	0.61	1.66	W
3hhrA	T175	2	48.8	3	2	3	2.86	1.23	W
3hhrA	F176	1.9	6.4	0	0	0	-	0.48	U
3hhrA	R178	2.4	66.9	0	0	0	0.85	0.69	W
3hhrA	I179	0.8	31.7	0	0	0	1.06	0.83	W
3hhrA	R183	0.5	0.3	0	0	0	-	-0.08	-
3hhrA	C182	1.01	26.7	0	0	0	-0.10	0.32	U
3hhrA	E186	0	5.2	0	0	0	-	0.06	-
3hhrB	R43	2.2	28.8	2	1	2	3.18	2.67	-
3hhrB	E44	1.8	28.5	3	3	3	0.05	0.02	-
3hhrB	N72	0.2	9.3	0	0	0	-	-0.12	-
3hhrB	W76	0.6	56	2	0	-1	2.0	2.49	-
3hhrB	T77	-0.25	24.4	3	2	2	0.93	1.02	-
3hhrB	W80	0	0.3	0	0	0	-	0.03	-
3hhrB	S98	-0.1	1	0	0	0	0.14	-0.01	-
3hhrB	S102	-0.2	25.2	0	0	0	0.33	0.96	-
3hhrB	I103	1.8	6.3	0	0	0	0.42	0.90	-
3hhrB	W104	4.5	139.2	7	6	10	4.98	5.17	-
3hhrB	I105	2	12.6	0	0	0	0.15	0.40	-
3hhrB	C108	0	10	0	0	0	-0.04	0.18	-
3hhrB	E120	-0.2	22.6	0	0	0	0.59	-0.29	-
3hhrB	K121	0.1	25.1	3	0	0	0.09	0.24	-
3hhrB	C122	0	37.9	1	1	1	-0.20	0.33	-
3hhrB	S124	0.2	11.2	0	0	0	-	-0.96	-
3hhrB	D126	1	26.9	0	0	0	-0.04	-0.95	-
3hhrB	E127	1	70	4	3	5	1.13	-1.54	-
3hhrB	D164	1.6	16.9	4	2	2	0.37	0.85	-
3hhrB	I165	2.2	10.2	0	0	0	0.06	0.15	-
3hhrB	Q166	0	42	0	0	0	0.42	0.29	-
3hhrB	K167	0	37.6	0	0	0	0.02	0.07	-
3hhrB	W169	4.5	87.8	7	7	13	3.47	4.44	-
3hhrB	Q216	0.9	5.9	0	0	0	-	0.02	-
3hhrB	R217	0.2	35.9	4	0	-1	0.17	-0.70	-
3hhrB	N218	0.3	107.5	4	0	0	1.48	0.00	-

Suffix added to the PDB entry code refers to chain name. Number added to residue symbol is the position of the residue in chain sequence.  $\Delta\Delta G$  = change of binding free energy upon mutation to alanine.  $\Delta ASA$  = change in water-accessible surface area of a residue upon dimer formation from monomer state. NA = number of atoms for a residue involved in inter-atomic sidechain - sidechain interactions.  $NA_i$  or  $NA_{ii}$  = number of side-chain atoms for a residue involved in favorable or unfavorable contacts.  $NC_i$  or  $NC_{ii}$  = number of favorable or unfavorable contacts. 'H' = predicted hot spot. 'W' = predicted warm residue. 'U' = predicted unimportant residue. '-' = the residue can not be identified.

**Supplementary Table 2: Protein-protein complexes with available 296 mutations at the interface**

PDB ID	Chain one	Chain two	No. mutations in chain one	No. mutations in chain two
1a4y	Angiogenin	Ribonuclease inhibitor	11	12
1ahw	Tissue Factor	Immunoglobulin Fab 5G9	7	-
1brs	Barnase	Barstar	7	5
1bxi	Colicin E9 immunity protein	Colicin E9 Dnase domain	17	-
1cbw	Basic Pancreatic Trypsin Inhibitor	Chymotrypsin	7	-
1dan	Blood coagulation factor Viia	Soluble Tissue Factor	24	34
1dfj	Ribonuclease A	Ribonuclease Inhibitor	-	14
1dvf	Fv D1.3	Fv E5.2	16	7
1fc2	Protein A	IgG Fc fragment	3	-
1gc1	Cd4	Envelope Protein Gp120	17	-
1jck	Staphylococcal enterotoxin C3	T cell antigen receptor V $\beta$	10	-
1vfb	Monoclonal antibody D1.3	Egg lysozyme	13	11
2ptc	Basic pancreatic trypsin inhibitor	Trypsin	1	-
3hfm	Lysozyme	IgG1 Fab (HyHEL-10)	13	2
3hhr	Human growth hormone	hGH receptor	39	26