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Discrete structural features among interface residue-level classes

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Abstract

Background: Protein-protein interaction (PPI) is essential for molecular functions in biological cells. Investigation on protein interfaces of known complexes is an important step towards deciphering the driving forces of PPIs. Each PPI complex is specific, sensitive and selective to binding. Therefore, we have estimated the relative difference in percentage of polar residues between surface and the interface for each complex in a non-redundant heterodimer dataset of 278 complexes to understand the predominant forces driving binding.

Results: Our analysis showed ~60% of protein complexes with surface polarity greater than interface polarity (designated as class A). However, a considerable number of complexes (~40%) have interface polarity greater than surface polarity, (designated as class B), with a significantly different p-value of 1.66E-45 from class A. Comprehensive analyses of protein complexes show that interface features such as interface area, interface polarity abundance, solvation free energy gain upon interface formation, binding energy and the percentage of interface charged residue abundance distinguish among class A and class B complexes, while electrostatic visualization maps also help differentiate interface classes among complexes.

Conclusions: Class A complexes are classical with abundant non-polar interactions at the interface; however class B complexes have abundant polar interactions at the interface, similar to protein surface characteristics. Five physicochemical interface features analyzed from the protein heterodimer dataset are discriminatory among the interface residue-level classes. These novel observations find application in developing residue-level models for protein-protein binding prediction, protein-protein docking studies and interface inhibitor design as drugs.

Background

Protein-protein binding is a known phenomenon in complex biological networks. The molecular principle of such binding is often elusive in nature. Understanding its driving forces using known protein complexes is essential. The analysis of existing protein-protein interaction (PPI) complexes from the Protein Data Bank (PDB) [1] is the key to gaining insights into recognition mechanisms and binding principles as reviewed elsewhere [2-6]. Sequence and structural investigations on the existing complexes has been carried out for several decades [3,7-10]. In these extensive surveys, structural features over diverse datasets

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of protein-protein complexes were typically averaged, obscuring information on individual proteins' structural integrity. Each individual complex is specific and sensitive to binding. Although, non-polar (or hydrophobic) interactions are known to play a major role in contributing to the driving force for binding, in a considerable number of complexes, polar interactions are also observed to contribute abundantly to the formation of a stable interface [11]. Therefore, it is often essential to study the relative difference in surface and interface polarity of each PPI complex to determine the major binding forces at the interface and determine their discriminatory features.

Interfaces are localized regions of surfaces with different physico-chemical properties compared to the rest of the surfaces, thereby driving binding to other molecules.



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Both physical and chemical features (including hydrophobicity, electrostatic interactions, binding energy, interface size, hydrogen bonds, salt bridges, disulphide bonds, planarity, sphericity, shape complementarity, amino acid chemical groups, and conserved residue clusters) govern the formation of protein interfaces as described elsewhere [7,9,12-18]. The chemical nature of residues forming a protein interface (amino acid residue composition) determines the hydrophobic effect of an interface. Non-polar (or hydrophobic) residues are observed to occur predominantly at the protein interface, playing a major role in contributing to the driving force for binding [7,13]. Interfaces are observed to be less non-polar (or hydrophobic) than the protein interior [13]. The residue composition of protein-protein interfaces was observed to be more similar to the protein surface than the protein interior [9].

Interfaces were observed to be significantly polar as well as non-polar with few charged groups, similar to the characteristics of the protein surface [12]. Structural analysis also revealed that charged and polar amino acids are involved in protein-protein interactions as reviewed elsewhere [19]. Charged and polar residues contributing to binding specificity and complex formation are demonstrated in a number of complexes such as human IL-4, human CD2 and CD58, barnase-barstar, Colicin E9, integrin $\alpha v\beta 6$ membrane protein and in intrinsically disordered proteins [20-25]. Shape complementarity, polar interactions, hydrogen bonding and salt bridges are also known to contribute to binding specificity and free energy of binding [17,24,26,27]. Also, charged and aromatic side chains are crucial for binding, determining the cation-pi, electrostatic and aromatic interactions [8]. The role of electrostatics in binding stability of protein-protein complexes is demonstrated [16]. These observations indicate that although PPIs are driven by non-polar interactions at the interface for a majority of complexes, in some cases polar interactions contribute to binding specificity (characteristic of polar residues) and likewise to complex stabilization. Therefore, a study on the relative percentage difference between surface and interface polarities of each protein complex is often essential. In our previous study, we have identified a class of complexes with more polar residues at the interface than the rest of the surface, where binding is mainly polar with a dataset of 198 complexes [11]. This observation has now been extended for an updated yet non-redundant dataset of 278 protein complexes to verify and identify any discriminatory features among these interface residue-level classes.

In this study, we have carried out a comprehensive structural analysis of 278 non-redundant heterodimeric protein complexes from the PDB. We estimated the relative difference in surface and interface polarities of each complex in the dataset, using percentage values of polar residues. This property divides the dataset into two interface classes as also observed in our previous study with a smaller dataset [11]. Class A has less polar residues at the interface than the rest of the surface (~60%) which is the 'classical' definition of a PPI complex and class B has more polar residues at the interface than the rest of the surface (~40%), is 'non-classical.' Therefore, we have investigated essential PPI structural features such as interface area (Δ ASA), the relative abundance of polar and non-polar residues at the interface (interface polarity abundance), hydrogen bonds (H-bonds), salt bridges, percentage of charged residues at the interface (interface charged residues%), solvation free energy gain upon interface formation ($\Delta^{i}G$), binding energy (BE), and electrostatics among these interface classes and their gleaned features are documented. We identified five key features (Δ ASA, interface polarity abundance, interface charged residues%, $\Delta^{i}G$ and BE) that are significantly different between the interface classes. These novel observations have implications for residue-level characterization of protein complexes to develop models for protein-protein binding prediction and docking studies.

Methods

Heterodimer dataset

We created a non-redundant heterodimer dataset of protein complexes from the PDB, using the RCSB PDB's advanced search interface. The following criteria were used for filtering: (i) resolution <= 3Å (ii) protein size >50 residues (iii) contains experimental data (iv) number of chains, entities and oligomeric state is set at 2 (v) devoid of DNA or RNA or a hybrid of such molecules with the protein or otherwise (vi) entries with mutations were not included and (vii) sequence identity cut-off is set to 30%, which is the minimum cut-off available in the PDB. As a second step, the USEARCH program [28] was used to further remove the redundancy among heterodimer complexes at sequence identity cut-off of 20%, as this threshold eliminates remote homology up to 25% sequence identity seen in structures as well [29].

Interface analysis

The interface of PPI complex is calculated as the change in solvent accessible surface area (Δ ASA) upon complex formation. The Surface Racer 5.0 program [30] was used to calculate ASA with a probe radius of 1.4Å and Lee and Richards implementation [31]. Interface residues with Δ ASA > 0.1Å² were considered for this analysis, as defined by Chakrabarti and Janin [32]. ASA was used to determine surface residues of each complex. The amount of polar, non-polar and charged residues at the interface was then estimated for the dataset. The interface polarity abundance (P%-NP%) is measured as the difference in the percentage of polar residues (P%) and percentage of non-polar residues (NP%) at the interface [11].

Classification based on relative interface-surface polarity

Interfaces are part of protein surface formed upon binding of individual subunits. Each protein complex has a specific composition of polar (P) and nonpolar (NP) residues at the surface (S) and at the interface (I). The distribution of polar and nonpolar residues at the interface of a protein complex describes the nature of the interface and the major driving force for binding. We have calculated the percentage of polar and nonpolar residues at the surface and interface for each complex in the dataset. Polar residues considered in the analysis are R, N, D, E, Q, H, K, S, T, and Y and non-polar residues are A, C, G, I, L, M, F, P, V, and W. Complexes were then grouped based on the relative difference in percentage of polar residues between surface (S) and the interface (I). Complexes with interface polarity less than the surface (represented as S>I) are grouped as class A, and those that have interface polarity greater than the surface (represented as S<I) are grouped as class B [11].

Intermolecular H-bonds and salt bridges calculation

We calculated the intermolecular hydrogen bonds for the dataset using HBPLUS program [33] at a distance of < 4Å. The H-bonds were extracted such that the donor and acceptor are from two different chains. Salt bridges were calculated using SBION program [34] within a distance of 4Å. The salt bridges were also extracted such that the oppositely charged atoms are from two different chains.

ΔⁱG and BE calculation

PDBePISA webserver [35] was used to obtain the solvation free energy gain upon interface formation (Δ^i G, in kcal/mol, with negative Δ^i G values indicating hydrophobic interface) and for the heterodimer dataset. BE values were calculated using the DCOMPLEX program [36] with the most negative value considered the strongest. The DCOMPLEX program uses DFIRE-based potentials [37] to calculate BE terms, without values for individual components (electrostatic, van der Waals, hydrophobic and entropic terms) contributing to BE. Initially, the program calculates the total atom-atom potential of mean force, G, for each protein structure as follows:

$$G = \frac{1}{2} \sum_{i,j} \left(\bar{u}_{i,j}, r_{i,j} \right)$$
(1)

where \bar{u} is the atom-atom potential of mean force between two atoms, i and j which are 'r' distance apart.

The total is over atomic pairs which are not from the same residue and a K factor is used to avoid double-counting of residue-residue and atom-atom interactions [36]. The binding energy between two interacting proteins A and B can be calculated as follows:

$$\Delta G_{bind} = G_{complex} (G_A + G_B) \tag{2}$$

where A and B are considered as two protein structures whose interface residues contribute most to ΔG_{bind} . Therefore, DCOMPLEX [36] uses the equation below to calculate BE:

$$\Delta G_{\text{bind}} = \frac{1}{2} \sum_{ij}^{interface} \left(\bar{u}_{i,j}, r_{i,j} \right)$$
(3)

Electrostatic potential at the interface

The surface electrostatic potential of chain A and chain B of a protein complex was calculated by solving Poisson-Boltzmann equation with dielectric constant (protein) of 4 using DEEPVIEW [38].

Statistical analysis

The Wilcoxon signed-rank test [39], a non-parametric statistical hypothesis test is used to compare the two interface classes to assess whether the mean ranks for the PPI features in the two classes differ (i.e. it is a paired difference test). The discriminatory PPI features among the two classes were thus tested for statistical significance with p < 0.05 (for the Wilcoxon signed-rank test) in RStudio [40].

Results and discussion

We calculated the amount of polar and non-polar residues at the surface and interface of each protein-protein complex and estimated their relative interface-surface polarities for classification into class A and class B (as described in Materials and Methods section), to determine the type of interactions predominantly driving protein-protein binding. Additional File 1: Table S1 shows the heterodimer dataset (278) divided into class A (165) and class B (113). Thus, 59.4% of complexes in our dataset belong to class A (relative surface polarity is greater than interface polarity), where non-polar interactions are predominant at the interface, as previously observed in a number of studies [7,13]. Nevertheless, 40.6% of complexes belong to class B (relative interface polarity is greater than surface polarity), where polar interactions are predominant at the interface, similar to the surface characteristics as also observed [12]. Class A and class B are significantly different with a p-value of 1.66E-45 (using Wilcoxon rank sum test) as shown in Additional File 2: Figure S1. Examples of class A and

class B complexes representing predominant non-polar and predominant polar interfaces (using the PDBsum [41] interaction analysis) respectively are shown in Figure 1.

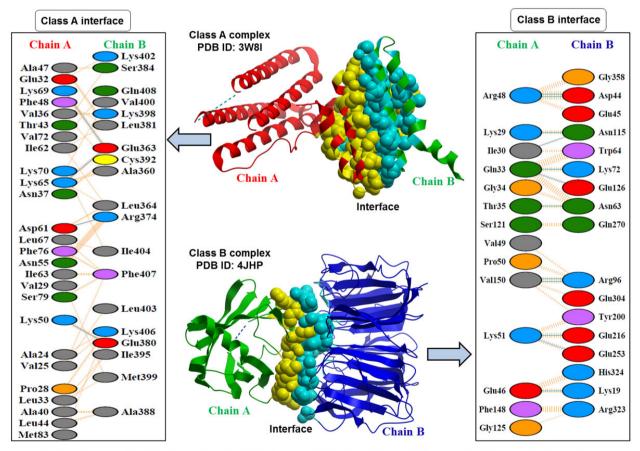
PPI features among class A and class B complexes

We carried out a statistical analysis of all the structural features (described in Materials and Methods section including Δ ASA, interface polarity abundance, interface charged residues%, H-bonds, salt bridges, Δ^i G, BE) in R program (using Wilcoxon rank sum test), to determine whether structural features discriminate among class A and class B complexes. Interestingly, five structural features showed significant difference among the interface classes as shown in Figure 2, with p-value < 0.05 (Table 1). The q-value in Table 1 is the smallest False Discovery Rate (FDR) at which a particular class (class A or class B)

would stay on the discriminatory features table. This is not identical to the p-value, which is the smallest false positive rate (FPR) at which a class appears positive on the discriminatory features table. The p-value is much stricter than the q-value. An FDR of 5% (q-value <0.05) is acceptable, which is accepting 5% of erroneous single results, according to Wilcoxon test [39]. These structural features are presented below, along with sets of other correlated properties and electrostatics among classes.

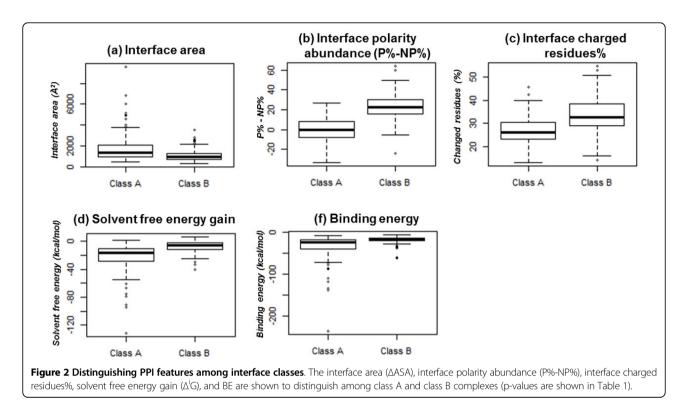
Interface polarity abundance among classes

Protein interfaces are composed of both polar and nonpolar residues. Some interfaces are abundant in nonpolar residues while few others are abundant in polar residues. The interface polarity abundance (P%-NP%) measure is significantly different among the interface classes with p = 7.01E-30 (Table 1 and Figure 2).



Residue colours: Positive (H,K,R); negative (D,E); S,T,N,Q = neutral; A,V,L,I,M = aliphatic; F,Y,W = aromatic; P,G = Pro & Gly; C = cysteine.

Figure 1 Examples of PPI interfaces in class A and class B complexes. The PDBsum [41] interaction analysis represents interaction residues on either chain with residues shown in different colours based on their properties and the coloured lines joining these residues representing the type of interaction between these residues. Class A complex shows a surface polariy of 60.28% and interface polarity of 37.84% (S>I) implying relatively less polar interactions at the interface (or relative abundance of non-polar interactions at the interface). Class B complex shows surface polariy of 50.69% and an interface polarity of 73.21% (S<I) implying relative abundance of polar interactions at the interface (or relatively less non-polar interactions at the interface).



ΔⁱG among classes

The solvation free energy gain upon interface formation $(\Delta^i G)$ is a measure of the interface stability in protein complexes [35]. The $\Delta^i G$ values are significantly different among interface classes with p = 7.43E-18 (Table 1) as shown in Figure 2.

BE among classes

The strength of binding among class A and class B complexes is estimated as a measure of BE in kcal/mol. The BE values are relatively stronger for class A complexes (average BE is -33.99 kcal/mol), as compared to class B complexes (average BE is -17.94 kcal/mol). The BE values are significantly different among interface classes with p = 2.63E-14 (Table 1) as shown in Figure 2.

Interface charged residues among classes

The percentage of charged residues at the interface is estimated for both classes. The interface charged residues% is

Table 1. PPI features distinguishing class A and class B (using Wilcoxon rank sum test)

PPI features	P-value	Q-value
Interface polarity abundance (P%-NP%)	7.01E-30	1.19E-28
Solvent free energy gain (Δ^i G),	7.43E-18	1.19E-16
Binding energy	2.63E-14	3.68E-13
Interface charged residues%	6.58E-13	8.55E-12
Interface area	1.31E-08	1.57E-07

significantly different among interface classes with p = 6.58E-13 (Table 1) as shown in Figure 2.

∆ASA among classes

The interface area (Δ ASA) of a complex is an important structural characteristic of PPI. We observed that class A complexes demonstrate comparatively larger interfaces than class B complexes. The Δ ASA is significantly different among the classes with p = 1.31E-08 (Table 1 and Figure 2).

Other correlations of interface features among classes

The stability of protein-protein binding depends on the number of hydrogen bonds and salt bridges formed between the two interacting subunits. Class A complexes show high correlation between intermolecular H-bonds and interface area (r = 0.9) as previously observed [7,42]. However, class B complexes alone show reduced trends (r = 0.73) between intermolecular H-bonds and interface area (Additional File 3: Figure S2), indicating that low quality of intermolecular hydrogen bonds is a characteristic of the large number of polar or charged residues across protein interfaces as previously observed [17]. Although salt bridges showed no distinguishing trends among classes, we observed that class B complexes are rich in salt bridges (average number of salt bridges is 6.5), as compared to class A complexes (average number of salt bridges is 5.8).

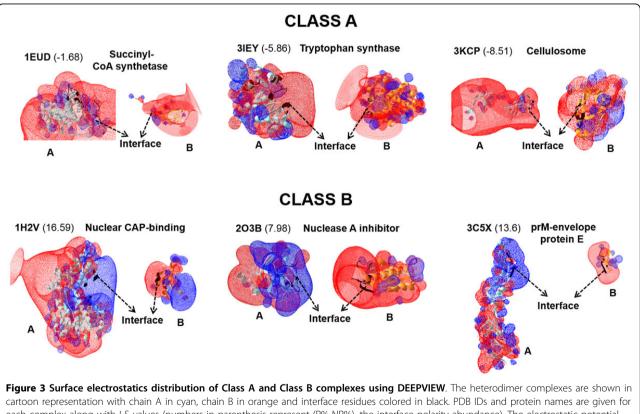
The BE values are proportional to interface area in the dataset (r = 0.96, shown in Additional File 4: **Figure S3**) as previously observed [43]. The Δ^{i} G values show relatively less correlation with interface area in class B complexes (r = -0.62) as compared to class A complexes (r = -0.92, Additional File 5: **Figure S4**). Moreover, the Δ^{i} G and BE is highly correlated among the dataset (r = 0.88) and class A (r = 0.91), however shows limited correlation among class B (r = 0.55, Additional File 6: **Figure S5**).

Electrostatic visualization maps among protein interface classes

We have studied the surface electrostatic potential solving Poisson-Boltzmann equation using DEEPVIEW for a few examples of class A and class B complexes. This shows common surface electrostatics at work amongst the class A and amongst the class B complexes. Interestingly, the class A complexes demonstrate similar distribution of charges at the protein interfaces of both chains, suggesting electrostatic energy may not contribute to binding energy among class A complexes. However, class B complexes show opposite charge distributions at the protein interfaces, suggesting electrostatic energy plays an important role in PPIs among class B complexes as shown in Figure 3. Therefore, the surface electrostatic potential maps give quick visual clues for identifying or distinguishing class A and class B complexes.

Conclusions

Structural analyses of known protein interfaces help in understanding the molecular principals of PPIs. Therefore, a comprehensive analysis of known structural interfaces of 278 complexes was carried out and their gleaned features are documented in this study. It is realized that each complex type is unique, specific and sensitive to binding. Nonetheless, there is a considerable degree of observed pattern among protein interface classes. We report two classes of interfaces, one class with less polar residues and the other class with more polar residues compared to the surfaces in bound state. The surfaces of proteins are quite polar and therefore, it is perhaps not surprising that some interfaces are polar as well and that PPI complex forms due to interactions among charged and polar residues. Thus, the need for a residue-level characterization of the interface is crucial



cartoon representation with chain A in cyan, chain B in orange and interface residues colored in black. PDB IDs and protein names are given to each complex along with I-S values (numbers in parenthesis represent (P%-NP%), the interface polarity abundance). The electrostatic potential images of class A complexes show that the interface of chain A and of chain B have same charges (similar colors), suggesting electrostatic energy may not favor protein binding in class A complexes. The electrostatic potential images of class B complexes show that the interface of chain A and chain B have opposite charges (different colors); suggesting electrostatic energy favors protein binding in class B complexes. in addition to other structural features. We document five discriminatory features (interface area, interface polarity abundance (P%-NP%), interface charged residues%, solvent free energy gain upon interface formation $(\Delta^{i}G)$, and binding energy) among the interface residuelevel classes. This is a first attempt towards classifying the complexes based on interface residue-level classes for the characterization of PPI features amongst these classes. These observations corroborate the need for classification of complexes in determining their combinatorial features and drawing consensus for common patterns in protein-protein recognition. These results provide molecular insights for protein-protein binding towards the development of residue-level prediction models in future studies. Additionally, mutation experiments using hot spot residue databases [44] and detailed interface residue characterization (cation-pi, electrostatic and aromatic interactions [8]) will further strengthen this study, for individual structures. Furthermore, extending this analysis for a larger dataset with a combined formulation of atomic and residue level features in future studies may improve protein-protein docking.

Additional material

Additional file 1: Table S1: Heterodimer dataset (278) divided into interface classes based on residue level relative surface-interface polarity. The PDB code is shown along with the specific chains used in this study.

Additional file 2: Figure S1: Class A and Class B are significantly different. The boxplot depicts class A and class B significantly different with a p-value of 1.66E-45 (using Wilcoxon rank sum test).

Additional file 3: Figure S2: Intermolecular H-bonds shows relatively low correlation with interface area in class B. Hydrogen bonds at the protein interface are highly correlated to interface area in the dataset (r = 0.88) and class A (r = 0.9), however shows relatively lower trends (r = 0.73) in class B.

Additional file 4: Figure S3: Binding energy is highly correlated to interface area. BEs at the protein interfaces are highly correlated to interface area with r = -0.96.

Additional file 5: Figure S4: Solvation free energy gain upon interface formation ($\Delta^{I}G$) shows limited correlation with interface area in class B complexes. $\Delta^{I}G$ shows high correlation with interface area in (a) heterodimer dataset (r = -0.88), and (b) class A (r = -0.92), however shows limited correlation in (c) class B complexes (r = -0.62).

Additional file 6: Figure S5: BE shows limited correlated with $\Delta^i G$ in class B. Binding energies at the protein interfaces are highly correlated to solvation free energy gain upon interface formation ($\Delta^i G$) in the dataset (r = 0.88) and class A (r = 0.91), however shows limited correlation between BE and $\Delta^i G$ in class B (r = 0.55).

Abbreviations

PPI, Protein-Protein Interaction; $\Delta^l G$, solvent free energy gain upon interface formation; BE, Binding Energy; H-bonds, Hydrogen bonds; PDB, Protein Data Bank; Δ ASA, interface area; P, Polar residues; NP, Non-polar residues; S, Surface polarity; I, Interface Polarity.

Competing interests

The authors declare that they have no competing interests.

Authors' contributions

Conceived and designed the experiment: GS. Data collected, analyzed and drafted the manuscript: GS. Interpretation of results and finalizing the manuscript: GS and SR. All authors read and approved the final manuscript.

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References

- Berman HM, Westbrook J, Feng Z, Gilliland G, Bhat TN, Weissig H, Shindyalov IN, Bourne PE: The Protein Data Bank. Nucleic Acids Res 2000, 28(1):235-242.
- Reichmann D, Rahat O, Cohen M, Neuvirth H, Schreiber G: The molecular architecture of protein-protein binding sites. *Curr Opin Struct Biol* 2007, 17(1):67-76.
- Keskin O, Gursoy A, Ma B, Nussinov R: Principles of protein-protein interactions: what are the preferred ways for proteins to interact? Chem Rev 2008, 108(4):1225-1244.
- Sudha G, Nussinov R, Srinivasan N: An overview of recent advances in structural bioinformatics of protein-protein interactions and a guide to their principles. Prog Biophys Mol Biol 2014.
- 5. Sowmya G, Ranganathan S: Protein-protein interactions and prediction: a comprehensive overview. *Protein Pept Lett* 2014, **21(8)**:779-789.
- Bahadur RP, Zacharias M: The interface of protein-protein complexes: analysis of contacts and prediction of interactions. *Cell Mol Life Sci* 2008, 65(7-8):1059-1072.
- Jones S, Thornton JM: Protein-protein interactions: a review of protein dimer structures. Prog Biophys Mol Biol 1995, 63(1):31-65.
- Gromiha MM, Yokota K, Fukui K: Energy based approach for understanding the recognition mechanism in protein-protein complexes. *Mol Biosyst* 2009, 5(12):1779-1786.
- Jones S, Thornton JM: Principles of protein-protein interactions. Proc Natl Acad Sci USA 1996, 93(1):13-20.
- 10. Gilson MK, Zhou HX: Calculation of protein-ligand binding affinities. Annu Rev Biophys Biomol Struct 2007, 36:21-42.
- 11. Sowmya G, Anita S, Kangueane P: Insights from the structural analysis of protein heterodimer interfaces. *Bioinformation* 2011, 6(4):137-143.
- 12. Lo Conte L, Chothia C, Janin J: The atomic structure of protein-protein recognition sites. *Journal of molecular biology* 1999, **285(5)**:2177-2198.
- Chothia C, Janin J: Principles of protein-protein recognition. Nature 1975, 256(5520):705-708.
- Miller S, Lesk AM, Janin J, Chothia C: The accessible surface area and stability of oligomeric proteins. *Nature* 1987, 328(6133):834-836.
- Saha RP, Bahadur RP, Pal A, Mandal S, Chakrabarti P: ProFace: a server for the analysis of the physicochemical features of protein-protein interfaces. *BMC Struct Biol* 2006, 6:11.
- 16. Dong F, Zhou HX: Electrostatic contribution to the binding stability of protein-protein complexes. *Proteins* 2006, 65(1):87-102.
- 17. Xu D, Tsai CJ, Nussinov R: Hydrogen bonds and salt bridges across protein-protein interfaces. *Protein Eng* 1997, **10**(9):999-1012.
- Guharoy M, Chakrabarti P: Conserved residue clusters at protein-protein interfaces and their use in binding site identification. *BMC Bioinformatics* 2010, 11:286.
- Sheinerman FB, Norel R, Honig B: Electrostatic aspects of protein-protein interactions. Curr Opin Struct Biol 2000, 10(2):153-159.
- 20. Wang T, Tomic S, Gabdoulline RR, Wade RC: How optimal are the binding energetics of barnase and barstar? *Biophys J* 2004, **87(3)**:1618-1630.

- Wang JH, Smolyar A, Tan K, Liu JH, Kim M, Sun ZY, Wagner G, Reinherz EL: Structure of a heterophilic adhesion complex between the human CD2 and CD58 (LFA-3) counterreceptors. *Cell* 1999, 97(6):791-803.
- Zhang JL, Simeonowa I, Wang Y, Sebald W: The high-affinity interaction of human IL-4 and the receptor alpha chain is constituted by two independent binding clusters. J Mol Biol 2002, 315(3):399-407.
- Kuhlmann UC, Pommer AJ, Moore GR, James R, Kleanthous C: Specificity in protein-protein interactions: the structural basis for dual recognition in endonuclease colicin-immunity protein complexes. J Mol Biol 2000, 301(5):1163-1178.
- Wong ET, Na D, Gsponer J: On the importance of polar interactions for complexes containing intrinsically disordered proteins. *PLoS Comput Biol* 2013, 9(8):e1003192.
- Sowmya G, Khan JM, Anand S, Ahn SB, Baker MS, Ranganathan S: A site for direct integrin alphavbeta6.uPAR interaction from structural modelling and docking. J Struct Biol 2014, 185(3):327-335.
- McCoy AJ, Chandana Epa V, Colman PM: Electrostatic complementarity at protein/protein interfaces. J Mol Biol 1997, 268(2):570-584.
- 27. Kundrotas PJ, Alexov E: Electrostatic properties of protein-protein complexes. *Biophys J* 2006, **91(5)**:1724-1736.
- Edgar RC: Search and clustering orders of magnitude faster than BLAST. Bioinformatics 2010, 26(19):2460-2461.
- Sander C, Schneider R: Database of homology-derived protein structures and the structural meaning of sequence alignment. *Proteins* 1991, 9(1):56-68.
- Tsodikov OV, Record MT, Sergeev YV: Novel computer program for fast exact calculation of accessible and molecular surface areas and average surface curvature. J Comput Chem 2002, 23(6):600-609.
- Lee B, Richards FM: The interpretation of protein structures: estimation of static accessibility. J Mol Biol 1971, 55(3):379-400.
- Chakrabarti P, Janin J: Dissecting protein-protein recognition sites. Proteins 2002, 47(3):334-343.
- McDonald IK, Thornton JM: Satisfying hydrogen bonding potential in proteins. J Mol Biol 1994, 238(5):777-793.
- Gupta PS, Mondal S, Mondal B, Islam RN, Banerjee S, Bandyopadhyay AK: SBION: A Program for Analyses of Salt-Bridges from Multiple Structure Files. *Bioinformation* 2014, 10(3):164-166.
- Krissinel E, Henrick K: Inference of macromolecular assemblies from crystalline state. J Mol Biol 2007, 372(3):774-797.
- Liu S, Zhang C, Zhou H, Zhou Y: A physical reference state unifies the structure-derived potential of mean force for protein folding and binding. *Proteins* 2004, 56(1):93-101.
- Zhou H, Zhou Y: Distance-scaled, finite ideal-gas reference state improves structure-derived potentials of mean force for structure selection and stability prediction. *Protein Sci* 2002, 11(11):2714-2726.
- Kaplan W, Littlejohn TG: Swiss-PDB Viewer (Deep View). Brief Bioinform 2001, 2(2):195-197.
- McDonald JH: Handbook of Biological Statistics. 3 edition. Baltimore, Maryland: Sparky House Publishing; 2014.
- RStudio: RStudio: Integrated development environment for R. v0.98.1091 edition. Boston; 2012.
- Laskowski RA: PDBsum new things. Nucleic Acids Res 2009, 37(Database): D355-359.
- 42. Gao Y, Wang R, Lai L: Structure-based method for analyzing proteinprotein interfaces. J Mol Model 2004, 10(1):44-54.
- Chen J, Sawyer N, Regan L: Protein-protein interactions: General trends in the relationship between binding affinity and interfacial buried surface area. Protein Sci 2013, 22(4):510-515.
- Ahmad S, Keskin O, Mizuguchi K, Sarai A, Nussinov R: CCRXP: exploring clusters of conserved residues in protein structures. *Nucleic Acids Res* 2010, 38(Web Server):W398-401.

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