

RESEARCH ARTICLE

Elucidating important structural features for the binding affinity of spike - SARS-CoV-2 neutralizing antibody complexes

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Abstract

The coronavirus disease 2019 (COVID-19) has affected the lives of millions of people around the world. In an effort to develop therapeutic interventions and control the pandemic, scientists have isolated several neutralizing antibodies against SARS-CoV-2 from the vaccinated and convalescent individuals. These antibodies can be explored further to understand SARS-CoV-2 specific antigen-antibody interactions and biophysical parameters related to binding affinity, which can be utilized to engineer more potent antibodies for current and emerging SARS-CoV-2 variants. In the present study, we have analyzed the interface between spike protein of SARS-CoV-2 and neutralizing antibodies in terms of amino acid residue propensity, pair preference, and atomic interaction energy. We observed that Tyr residues containing contacts are highly preferred and energetically favorable at the interface of spike protein-antibody complexes. We have also developed a regression model to relate the experimental binding affinity for antibodies using structural features, which showed a correlation of 0.93. Moreover, several mutations at the spike protein-antibody interface were identified, which may lead to immune escape (epitope residues) and improved affinity (paratope residues) in current/emerging variants. Overall, the work provides insights into spike protein-antibody interactions, structural parameters related to binding affinity and mutational effects on binding affinity change, which can be helpful to develop better therapeutics against COVID-19.

KEYWORDS

binding affinity, COVID-19, mutational analysis, neutralizing antibodies, regression analysis, SARS-CoV-2

1 | INTRODUCTION

Coronavirus disease 2019 (COVID-19) has caused severe global economic depression and the loss of millions of human lives. As of November 2021, there have been more than 246 million confirmed cases and over 5 million deaths due to SARS-CoV-2 infection (<https://covid19.who.int/>). The causative agent of the COVID-19 pandemic, SARS-CoV-2, is a single-stranded RNA virus that is around 30 000 base pairs in length. SARS-CoV-2 infection in humans is initiated by the interaction of the receptor binding

domain (RBD), present in S1 subunit of the spike (S) protein, with the human ACE2 receptor leading entry to the host cells.¹ Researchers have identified several neutralizing antibodies induced against SARS-CoV-2, either by vaccines or natural infection. The major portion of these neutralizing antibodies controls viral infection by blocking the binding of SARS-CoV-2 to the human ACE2 receptor.² In other neutralization mechanisms, antibodies may bind the N-terminal domain in the S1 subunit to potentially restrain the conformational changes of the S protein^{3,4} or bind S2 to inhibit membrane fusion and subsequent viral entry.⁵

An in depth analysis of virus-host interaction is essential to understand the pathophysiology of infection and to develop therapeutic intervention.⁶ The studies on SARS-CoV-2 protein interaction networks have revealed various drug targets for drug discovery.^{7,8} The deep mutational scanning study by Greaney et al.⁹ assessed all possible single amino acid variants in the spike protein and provided immune escape maps for mutations in the presence of antibodies. They applied these maps to five potent neutralizing antibodies (COV2-2094, COV2-2165, COV2-2479, COV2-2050, and COV2-2499) isolated from SARS-CoV-2 convalescent patients and found K378E, K378N, E484K, G446D, and Q498R as immune escape mutations. Another study detected variants that escaped either neutralizing SARS-CoV-2 mAbs or convalescent plasma and observed that mutations S477N and E484K rank prominently among mAb escape mutations.¹⁰ Although these mutational studies have significantly improved our knowledge on immune escape variants, they lack a large-scale analysis to understand the biophysical factors leading to decreased/loss of binding. The analysis of SARS-CoV-2 neutralizing antibodies would provide better insights, allowing engineering of more potent antibodies for current/emerging strains.¹¹⁻¹³

In this work, we have analyzed the interface of SARS-CoV-2 spike protein and neutralizing antibody complexes with respect to ACE2 binding to understand the residue propensity, pair preference, and atomic interaction energy. Tyr residues show the highest propensity, and the YY pairs have highest pair preference at the interface of spike-antibody complexes. Further, we use structural features including atomic interactions, interface area, surrounding hydrophobicity, hydrophobic interactions, aromatic-aromatic interactions to develop a regression model to relate the binding affinity (K_D) for each antibody. The comprehensive mutational study of the residues at the spike-antibody interface revealed the important sites in spike protein, which lead to decreased binding affinity. The study is in agreement with experimental studies performed on selected neutralizing antibodies, C121, C144, and C135.¹⁴ We extended the analysis to mutate the residues at the antibody interface to identify mutations that may improve the affinity of neutralizing antibodies. The analysis will be helpful to (a) understand the interactions between the spike protein and the neutralizing antibodies, (b) develop better therapeutics against COVID-19 based on interaction preferences, (c) understand immune escape for emerging variants of SARS-CoV-2, and (d) improve stability and binding affinity of neutralizing antibodies.

2 | MATERIALS AND METHODS

2.1 | Dataset preparation

2.1.1 | Construction of a dataset for spike-neutralizing antibody complex structures

The experimentally reported structures of 77 SARS-CoV-2 neutralizing antibodies in complex with spike protein were obtained from the literature. The dataset was further screened based on (a) structure

resolution (highest resolution among available structures for particular antibody complexes), (b) single antibody per structure, and (c) sequence identity (<95% for whole sequence and <80% for CDRH3 region). The final antibody dataset contained 29 spike protein-antibody complexes (Table 1). The ACE2 bound spike protein structure (PDB id: 6MOJ) was used as a reference for comparison.

The residues at the interface of spike protein and antibody were identified using all heavy atom distance cutoff of 4 Å. For structural analysis, we removed heteroatom coordinates from the PDB files and retained one subunit of the spike protein RBD and antibody (heavy and light chain). A set of 983 heterodimer complexes were collected from the PDB database¹⁵ and used as a reference to compare the

TABLE 1 K_D values for neutralizing antibodies and RBD-ACE2 complex

S. no.	PDB ID	Antibody	K_D (nM)	ΔG (kcal/mol)
1	6WPT	S309	<0.001	-16.3
2	6XC2	CC12.1	17	-10.55
3	6XC4	CC12.3	14	-10.67
4	6XCN	C105	14	-10.67
5	6XE1	CV30	3.63	-11.47
6	6XEY	Fab2-4	N/A	N/A
7	6XKP	CV07-270	N/A	N/A
8	6XKQ	CV07-250	0.04	-14.14
9	7B3O	STE90-C11	8.1	-10.99
10	7BWJ	P2B-2F6	5.14	-11.26
11	7BYR	BD23	N/A	N/A
12	7BZ5	B38	70.1	-9.72
13	7C01	CB6	2.49	-11.69
14	7CAK	H014	0.1	-13.6
15	7CHB	BD-236	2.8	-11.61
16	7CHH	BD-368-2	0.54	-12.59
17	7CJF	P4A1	1.78	-11.89
18	7CWO	P17	N/A	N/A
19	7JMW	COVA1-16	0.2	-13.18
20	7JV2	S2H13	N/A	N/A
21	7K43	S2-M11	66	-9.75
22	7K45	S2-E12	1.6	-11.95
23	7K8U	C002	11	-10.81
24	7K8V	C110	1.3	-12.07
25	7K8W	C119	10	-10.87
26	7K8Y	C121	0.5	-12.63
27	7K8Z	C135	6	-11.17
28	7K90	C144	18	-10.52
29	7KFV	C1A-B3	76.3	-9.67
30	6MOJ	RBD-ACE2	14.95 ^a	-10.65

Note: N/A, data are not available.

^aAverage of 14.7 and 15.2 nM.

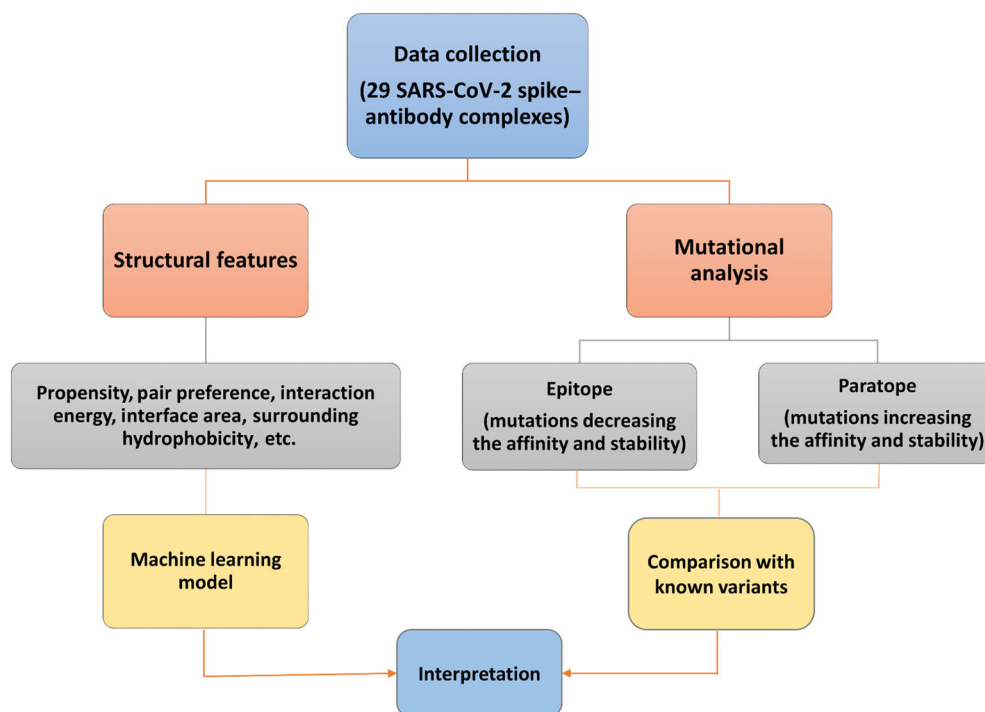


FIGURE 1 Overview of the workflow

propensities of interface residues. An overview of the study is represented in Figure 1.

2.1.2 | Binding affinity of spike-neutralizing antibodies

The experimental binding affinities (K_D) were obtained from the literature for 24 spike-antibody complexes. The binding free energy (ΔG) was calculated from K_D using the following equation:

$$\Delta G = -RT \ln(1/K_D) \quad (1)$$

where, $R = 8.31 \text{ J/mol/K}$, is the gas constant and T is the temperature (298 K). The experimental ΔG values are presented in Table 1.

2.2 | Computational procedures

2.2.1 | Binding propensity

We have calculated the propensity of 20 amino acids to be present at the interface ($P_{\text{interface}}$) using the following equation¹⁶:

$$P_{\text{interface}}(i) = \frac{\% \text{ of residue } i \text{ in interface}}{\% \text{ of all residues in interface}} = \frac{n_{\text{interface}}(i)/N(i)}{n_{\text{interface}}/N} \quad (2)$$

where, $n_{\text{interface}}(i)$ and $N(i)$ are the number of residues of type i at the interface and protein, respectively. $n_{\text{interface}}$ is the number of residues at the interface, and N is the total number of residues in the protein.

2.2.2 | Residue pair preference and interaction energy

The pair preference ($\text{Pair}[i,j]$) for the spike-antibody residue pairs at the interface was computed using the following equation¹⁷:

$$\text{Pair}(i,j) = \sum N_{ij} \times 100 / \left(\sum N_i \times \sum N_j \right) \quad (3)$$

where, i and j stand for the interface residues in spike and antibodies, respectively. N_{ij} is the number of interacting residues of type i in spike and j in antibodies. $\sum N_i$ and $\sum N_j$ are the total number of residues of type i in spike and j in antibodies, respectively.

The atomic interaction free energy (E_{inter}) between interface residues of the spike and antibody was calculated using AMBER potential.¹⁸ It is given as Equation (4)¹⁷:

$$E_{\text{inter}} = \sum \left[\left(A_{ij}/r_{ij}^{12} - B_{ij}/r_{ij}^6 \right) + q_i q_j / \epsilon r_{ij} \right] \quad (4)$$

where, $A_{ij} = \epsilon_{ij}^* (R_{ij}^*)^{12}$ and $B_{ij} = 2\epsilon_{ij}^* (R_{ij}^*)^6$; $R_{ij}^* = (R_i^* + R_j^*)$ and $\epsilon_{ij}^* = (\epsilon_i^* \epsilon_j^*)^{1/2}$; R^* and ϵ^* are, respectively, the van der Waals radius and well depth, q_i and q_j are, respectively, the charges for the atoms i in spike and j in antibodies, and r_{ij} is the distance between them.

2.2.3 | Computation of interface area

Accessible surface areas of the RBD region in the spike protein (S_E), antibody (S_{H+L}) as well as the entire complex (S_{complex}) were calculated

by rolling a water molecule of radius 1.4 Å on the protein/complex surface as described in our previous study.¹⁹ Further, the interface area ($S_{\text{interface}}$) was calculated using Equation (5).

$$S_{\text{interface}} = (S_E + S_{H+L}) - S_{\text{complex}} \quad (5)$$

2.2.4 | Computation of surrounding hydrophobicity

The surrounding hydrophobicity of residues in spike protein–antibody complexes was computed using the following equation.²⁰

$$H_p(i) = \sum_{j=1}^{20} n_{ij} h_j \quad (6)$$

where, $H_p(i)$ is surrounding hydrophobicity of i th residue of the protein. n_{ij} is the total number of surrounding residues of type j around residue i within 8 Å distance (between C_α atoms). h_j is hydrophobicity value for the residue type j (in kcal/mol) obtained from thermodynamic transfer experiments.^{21,22} The average hydrophobicity indices (H_p) was calculated for the interface residues in each spike–antibody complex using PDBparam.²³

2.3 | Development of regression model

2.3.1 | Features used in the regression model

In addition, we computed the interaction energy and number of contacting residues using PRODIGY,²⁴ and various types of atomic interactions from PIC (Protein Interactions Calculator) server.²⁵ The details of all the features for the spike–neutralizing complexes considered in the present study are presented in Table S1.

2.3.2 | Development of multiple regression models

We have developed multiple regression equations to relate structure-based features with binding affinity of spike–antibody complexes. It is defined as,

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_p x_{ip} + \epsilon \quad (7)$$

where, i is the number of observations, y_i is a dependent variable (binding affinity), x_i are structure-based parameters, $\beta_0, \beta_1, \dots, \beta_p$ are regression coefficients, and ϵ is the error term of the model.

A systematic forward feature selection approach was utilized to select the optimum number of features with the best performance. The regression model was evaluated by the Pearson correlation coefficient to measure the strength of the relationship between two variables and the mean absolute error (MAE) to examine the absolute

difference between predicted and experimental affinity values. The model was further validated using a jackknife test, where regression equations were developed to predict ΔG using $(n - 1)$ data points to predict the performance on n th datapoint, recursively.

2.4 | Mutational scanning based on change in affinity and stability

Change in affinity and stability upon mutation of the interface residues (paratope and epitope) for all antibody complexes were obtained using sequence and structure based methods. The ProAffiMuSeq server predicts protein–protein binding affinity change upon mutation using sequence-based features and functional class.²⁶ mCSM relies on graph-based signatures to study missense mutations and predicts change in stability and affinity.²⁷ The CUPSAT prediction model uses amino acid–atom potentials and torsion angle distribution to assess the amino acid environment of the mutation site and predicts the change in stability upon mutation.²⁸ FoldX software uses empirical function to evaluate the effect of mutations on the stability, interaction, folding, and dynamics of proteins.²⁹ For FoldX, “RepairPDB” command was used to rectify pdb files, the residues present at the interface of spike protein and antibody were mutated systematically using “BuildModel” command, and the interaction energy was calculated using “AnalyseComplex” command, and stability was calculated as “Stability” command. The change in interaction energy was further calculated by subtracting the values of mutant with wild-type.³⁰

3 | RESULTS AND DISCUSSION

3.1 | Comparison of binding site residues in antibodies and ACE2 with spike protein

The SARS-CoV-2 binds with the human ACE2 at the receptor binding domain (RBD) of the spike protein. We compared the epitope regions in spike protein for each antibody and observed that four antibodies (COVA1-16, H014, S309, and C135) have different epitopes than the ACE2 binding site. COVA1-16 and H014 antibodies inhibit ACE-2 binding through steric hindrance^{31,32} whereas, S309 and C135 antibodies do not compete with ACE2 binding and recognize a proteoglycan epitope on the SARS-CoV-2 spike, distinct from the receptor-binding motif.^{4,33} Hence, for a fair comparison with similar binding regions, these four antibodies were not included in the interface analysis. Epitopes of the remaining 25 antibodies overlap with the ACE2 binding site.

3.2 | Amino acid propensity at the spike–antibody interface

To identify the residues that have a high frequency of occurrence at the spike–antibody interface, we calculated the propensity of

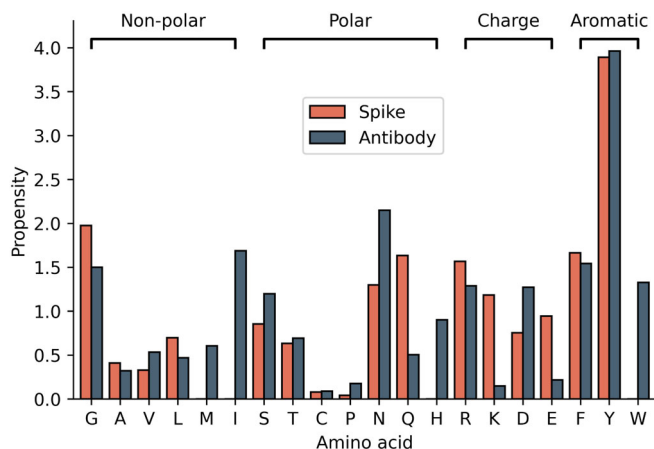


FIGURE 2 Propensity of interface residues for spike and antibody

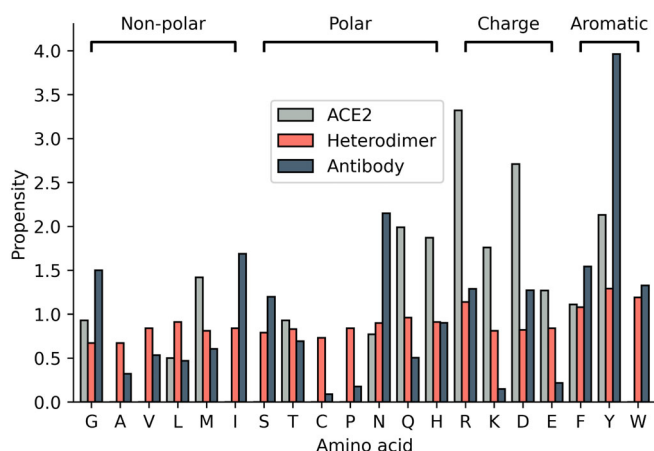


FIGURE 3 Binding propensity of amino acid residues at the interfaces of ACE2, heterodimers, and neutralizing antibodies

interface residues (Equation (2)). The propensity of Tyr was observed to be the highest at the interface of spike–antibody complexes (Figure 2). Interestingly, Tyr predominantly occurred at the interface of antigen–antibody³⁴ and other protein–protein complexes.³⁵ It has been reported that tyrosine residues are exceptionally versatile for mediating contacts at interfaces³⁶ as they can form stable aromatic interactions. Other residues with high propensity are Asn, Ile, Phe, and Gly from antibody interface and Gly, Phe, Gln and Arg from spike protein interface, respectively (Figure 2). On the other hand, Met, His, Trp, and Ile residues are absent at the spike interface, and Cys has the lowest propensity of 0.09 at the antibody interface. The residues which show a difference of greater than 1 in residue propensity (in spike and antibody) are Ile, Trp, Gln, and Lys (Figure 2).

Further comparison of the interface of antibodies with ACE2 showed that Arg has the highest propensity in ACE2 compared with Tyr in antibodies. Other residues in ACE2 with higher propensity include Asp, Tyr, Gln, and His (Figure 3). Similarly, a comparison of antibody and heterodimer protein interfaces showed that Tyr and Phe had high frequency in both interfaces (Figure 3). The Gly and Ser had a lower propensity in heterodimers and ACE2 but a higher propensity in antibodies. Cys and Ala had a lower propensity in all of the three

datasets (Figure 3), which agrees well with previous studies reported in the literature for ACE2 and heterodimer interfaces.^{37,38}

3.3 | Residue pair preference and interaction energy

The analysis of residue pair preference at the interface of the spike–antibody complexes showed that Tyr containing residue pairs (YY, FY, GY, YG, YS) dominate the topmost preferred pairs. These residue pairs are present in 70% of the antibodies in the dataset. These highly preferred pairs are also more energetically favorable (Table 2) indicating that aromatic interactions are important for binding of antibodies to the SARS-CoV-2 RBD. Along with aromatic–aromatic interactions, nonpolar– π and polar– π interactions are also preferred, and the polar– π interactions are energetically significant to protein folding and function.³⁹ A previous study on 200 antibody–antigen complexes has also shown that Tyr residues are preferred at both the epitope and paratope regions.⁴⁰ In our previous mutational study, we also observed that Tyr residue at the ACE2 binding regions of spike protein are highly conserved along with Gly residues in all ACE2-binding coronaviruses.⁴¹ Hence, Tyr residue interactions are potentially important for developing highly specific antibodies or small drug molecules against SARS-CoV-2.

Out of 20 topmost preferred pairs at the spike–antibody interface (Table S2) 17 contains hydrophobic residues (9 hydrophobic–hydrophobic interactions) and these hydrophobic residues play a vital role at protein–protein interfaces.^{42,43} It has been shown that the dominance of hydrophobic contacts between SARS-CoV-2 and ACE2 enhances the binding affinity of SARS-CoV-2 for ACE2.⁴⁴ In the heterodimer dataset, CC is the most preferred pair, followed by KD, WI, RD, and VF, which does not contain Tyr and is different from the spike–antibody interface. It is important to note that Cys has the lowest propensity at the neutralizing antibody interfaces, and therefore, such pairing is not preferred at the spike–antibody interface.

3.4 | Structural features influencing experimental binding affinity (K_D)

The binding free energy values for antibodies discussed in this work range from -9.7 to -16.3 kcal/mol (Table 1). The binding affinity values for spike–ACE2 complexes are reported as 4.7 ,¹ 14.7 ,⁴⁵ and 15.2 nM⁴⁶ in the literature. In comparison to ACE2 binding affinity (~ 15 nM), most of the antibodies showed greater binding affinities leading to effective competitive inhibition of SARS-CoV-2 (Table 1).

3.4.1 | Analysis of antibodies with experimental binding affinity

We also calculated the interface surface area and average surrounding hydrophobicity of the spike protein–ACE2 complex and spike–antibody complexes. The surface area of the interacting region for

TABLE 2 Energetically favorable pairs (<-0.2 kcal/mol) within topmost 50 preferred pairs

S. no.	Pair	Pair preference ^a	Energy (kcal/mol)	S. no.	Pair	Pair preference ^a	Energy (kcal/mol)
1	YY	11.69	-0.32	24	KD	2.30	-0.23
2	FY	5.84	-0.38	25	RY	2.28	-0.43
3	YI	5.75	-0.21	26	FD	2.20	-0.26
4	YN	5.60	-0.25	27	NG	2.19	-0.43
5	KY	4.70	-0.39	28	EY	1.91	-0.38
6	YR	4.61	-0.35	29	KS	1.81	-0.38
7	GY	4.03	-0.28	30	YL	1.80	-0.39
8	EH	3.91	-0.42	31	FF	1.78	-0.25
9	YG	3.73	-0.24	32	NF	1.77	-0.34
10	YS	3.72	-0.40	33	LY	1.65	-0.42
11	YF	3.60	-0.36	34	YD	1.65	-0.37
12	FR	3.42	-0.44	35	EW	1.60	-0.23
13	YW	3.30	-0.37	36	QN	1.54	-0.56
14	AN	3.28	-0.49	37	AF	1.49	-0.22
15	FM	2.67	-0.42	38	FW	1.47	-0.22
16	QW	2.54	-0.21	39	EN	1.46	-0.30
17	RN	2.54	-0.64	40	GG	1.42	-0.31
18	NR	2.42	-0.30	41	YT	1.38	-0.35
19	YM	2.40	-0.39	42	LM	1.35	-0.30
20	DY	2.40	-0.26	43	GF	1.33	-0.32
21	QY	2.39	-0.37	44	QS	1.32	-0.40
22	ER	2.33	-0.43	45	TF	1.26	-0.27
23	QM	2.32	-0.36	46	RS	1.25	-0.28

^aThe pair preference values are multiplied by 10^3 .

antibodies ranged between 370 and 1300 Å² on the spike protein. SARS-CoV-2 spike-ACE2 complex had a relatively larger surface area of 1860 Å² than antibodies. The contact surface area in the spike-antibody proteins shows a low positive correlation with enhanced binding affinity ($r = .13$), which is consistent with previous studies.⁴⁷ Similarly, the average surrounding hydrophobicity for the interacting region in antibodies ranged between 10.24 and 14.84 kcal/mol compared with 12.31 kcal/mol for ACE2. We have shown previously on coronaviruses that a higher hydrophobic environment at the interface can improve the binding of the complex.⁴¹ The surrounding hydrophobicity also correlates positively with the binding affinity ($r = .26$).

3.4.2 | Factors influencing binding free energy (ΔG)

The single property correlation of structural features to ΔG showed the highest negative correlation of -0.55 for charged-charged interfacial contacts and a highest positive correlation of 0.5 for aromatic-aromatic interactions, calculated within the distance of 4.5–7 Å.⁴⁸ A systematic forward feature selection approach was further utilized to select the optimum number of features with the best performance. We obtained a correlation of 0.93 and a mean absolute error (MAE) of 0.48 kcal/mol for the final regression model with three features

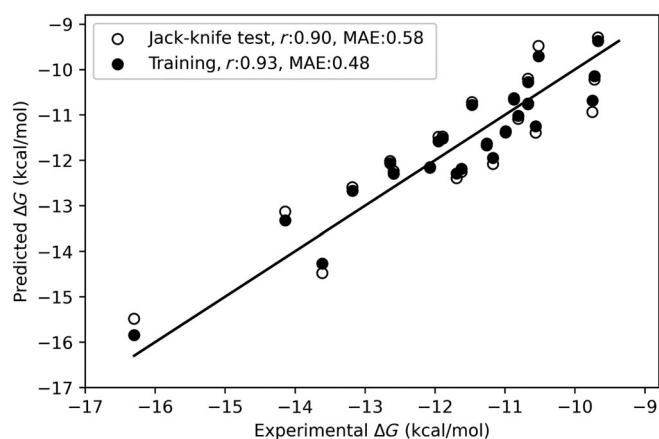


FIGURE 4 A scatter plot showing the relationship between experimental and predicted binding energies on training and jackknife test

(Figure 4). The features include total energy (E_t , in kcal/mol), aromatic-aromatic interactions ($I_{\text{aro-aro}}$), and number of nonpolar atomic contacts (A_{np}). The multiple regression equation obtained from the model for binding free energy (ΔG , in kcal/mol) is given below:

TABLE 3 Mutation sites affecting binding affinity and stability in the epitope(s) and paratope(s) of spike-antibody interface

Pdb id	Interface (chain)	Interface residues
6WPT	Paratope (L)	D93
	Epitope (C)	N334 , L335, P337, G339, E340, N343, A344, T345, R346, N354, K356, R357, S359, N360, C361, L441, R509
6XC2	Paratope	-
	Epitope (A)	R403 , D405, R408, T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, S459, N460, Y473, Q474, A475, G476, F486, N487, Y489, Q493, S494, Y495, G496, Q498, T500, N501, G502, Y505
6XC4	Paratope	-
	Epitope (A)	R403 , D405, T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, N460, Y473, A475, G476, S477, F486, N487, Y489, Y495, N501, Y505
6XCN	Paratope (H,L)	G54, S56, G29, Y30, K31
	Epitope (C)	D405 , T415, G416, K417, Y421, Y453, F456, R457, K458, N460, Y473, A475, G476, F486, N487, G502, Y505
6XE1	Paratope (H,L)	D97, S27A
	Epitope (E)	R403 , T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, N460, Y473, A475, G476, F486, N487, Y489, Q493, N501, G502, Y505
6XEY	Paratope (J)	A99
	Epitope (C)	Y449 , L455, F456, V483, E484, G485, F486, Y489, F490, L492, Q493, S494
6XKP	Paratope (H)	R96
	Epitope (A)	R346 , F347, S349, Y351, K444, V445, G446, G447, N448, Y449, N450, Y451, L452, T470, E484, F490, L492, Q493, S494, Q498
6XKQ	Paratope (H,L)	M100E, D101, S27A, A29
	Epitope (A)	R403 , G446, Y449, Y453, L455, F456, A475, G476, S477, T478, G485, F486, N487, Y489, Q493, Y495, Q498, N501, Y505
7B3O	Paratope (H,L)	N92, S31, R97, A100, D101
	Epitope (E)	R403 , D405, E406, R408, Q409, T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, S459, N460, Y473, Q474, A475, G476, F486, N487, Y489, Q493, S494, Y495, G496, Q498, T500, N501, G502, Y505
7BWJ	Paratope (L)	G31
	Epitope (E)	R346 , K444, G446, G447, N448, Y449, N450, L452, V483, E484, G485, F490, S494
7BYR	Paratope (H)	S31, Y32, T53, N54, D73, Q100, S103, W105
	Epitope (B)	G446, Y449, E484, G485, F486, Y489, F490, L492, Q493, G496, Q498, N501, Y505
7BZ5	Paratope	-
	Epitope (A)	R403 , D405, E406, Q409, T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, N460, Y473, Q474, A475, G476, E484, F486, N487, Y489, F490, L492, Q493, Y495, G496, Q498, T500, N501, G502, Y505
7C01	Paratope (H,L)	D104, S30
	Epitope (A)	R403 , D405, E406, R408, Q409, T415, G416, K417, D420, Y421, L455, F456, R457, K458, N460, Y473, Q474, A475, G476, S477, F486, N487, Y489, Q493, Y495, G502, Y505
7CAK	Paratope (D,E)	I2, S27, S30, N91, W93, D60, D102, Y105
	Epitope (A)	Y365 , Y369, A372, S373, F374, S375, T376, F377, K378, S383, P384, T385, R408, P412, G413, Q414, N437, V503
7CHB	Paratope (H,L)	G26, S31, A102, D106, Q27, S30
	Epitope (R)	R403 , T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, S459, N460, Y473, A475, G476, F486, N487, Y489, Q493, Y495, G496, Q498, T500, N501, G502, V503, G504, Y505
7CHH	Paratope (D)	S25, D106
	Epitope (A)	K444, Y449, N450, L452, N481, G482, V483, E484, G485, F490
7CJF	Paratope (A,B)	S30, F58, D106, S93
	Epitope (C)	R403 , D405, Q409, T415, G416, K417, D420, Y421, Y453, L455, F456, R457, K458, N460, Y473, Q474, A475, G476, S477, F486, N487, Y489, Q493, Y495, G496, Q498, T500, N501, G502, Y505
7CWO	Paratope (H)	S31
	Epitope (A)	L455 , T470, N481, G482, V483, E484, G485, F486, Y489, F490, L492, Q493

TABLE 3 (Continued)

Pdb id	Interface (chain)	Interface residues
7JMW	Paratope (H)	<i>P96</i>
	Epitope (A)	Y369 , <i>S371</i> , F377 , K378 , C379 , Y380 , G381 , V382 , S383 , P384 , T385 , R408 , P412 , G413 , Q414 , T415 , G416 , D427 , F429
7JV2	Paratope (H,L)	<i>Y102</i> , <i>R63</i>
	Epitope (A)	G446 , Y449 , N481 , G482 , V483 , <i>E484</i> , G485 , F486 , F490
7K43	Paratope (H)	G31 , <i>T74</i>
	Epitope (A)	G446 , Y449 , L452 , L455 , F456 , <i>E484</i> , G485 , F486 , Y489 , F490 , L492 , Q493 , S494 , G496
7K45	Paratope (H)	<i>G54</i> , <i>G104</i> , S105
	Epitope (B)	L455 , Y473 , A475 , G476 , S477 , G485 , F486 , N487 , C488 , Y489
7K8U	Paratope (H,L)	<i>G26</i> , <i>S30</i> , <i>A31</i> , Y92 , <i>G93</i> , <i>T95</i>
	Epitope (A)	K444 , V483 , F486 , F490
7K8V	Paratope (H)	<i>D99</i> , <i>V100D</i> , P100G
	Epitope (A)	T345 , <i>R346</i> , L441 , D442 , N448 , Y449 , N450 , L452 , F490 , Q498 , P499 , T500 , R509
7K8W	Paratope (H,L)	<i>S58</i> , <i>D100B</i> , Y100D , Y100E , K30 , <i>S92</i>
	Epitope (A)	<i>K444</i> , <i>V445</i> , G446 , Y449 , <i>N450</i> , E484 , Q493 , <i>S494</i> , <i>Q498</i> , Y505
7K8Y	Paratope (G)	T58 , T74 , G110
	Epitope (B)	K444 , Y449 , F456 , <i>E484</i> , G485 , F486 , N487 , Y489 , F490 , Q498
7K8Z	Paratope	–
	Epitope (A)	T345 , <i>R346</i> , S438 , N439 , N440 , P499
7K90	Paratope (H)	<i>G54</i> , <i>G55</i> , S56 , K73 , N76
	Epitope (B)	Y449 , L455 , F456 , <i>V483</i> , G485 , F486 , N487 , Y489 , F490 , Q493 , S494
7K9V	Paratope (H,L)	S98 , S67
	Epitope (A)	R403 , T415 , G416 , K417 , D420 , Y421 , Y453 , L455 , F456 , R457 , <i>K458</i> , S459 , N460 , Y473 , A475 , G476 , F486 , N487 , Y489 , Q493 , <i>S494</i> , Y495 , G496 , <i>Q498</i> , T500 , N501 , G502 , <i>G504</i> , Y505

Note: (i) The residues shown in italics for paratope increase the binding affinity, bold residues increase the stability, while residues shown for epitope decrease the binding affinity and stability. (ii) The residues are highlighted if two out of three methods satisfy the criteria that at least 50% of the mutations in each residue increases (for paratope) or decreases (for epitope) the affinity and stability. (iii) Notation for residues; Wild type residue followed by residue number.

$$\Delta G = 0.29 \times E_t + 0.38 \times I_{\text{aro-aro}} + 0.04 \times A_{np} - 10.78 \quad (8)$$

We have also performed a jack-knife test to validate the robustness of the model and observed the correlation of 0.90 and MAE of 0.58 between experimental and predicted binding energies (Figure 4). For comparison, we used PRODIGY (structure based method)²⁴ and PPA-Pred (sequence based method)⁴⁹ for predicting the binding affinity of the same set of 24 protein–protein complexes and we obtained an MAE of 3.05 and 2.64 kcal/mol, respectively. Further, we tested our model for five typical antibodies [CT-P59 (7CM4), P5A-3C12 (7DOB), P5A-1B9 (7CZX), Fab2-43 (7L56), and P5A-2F11 (7CZY)] that are not considered in our dataset, which showed an MAE of 1.15 kcal/mol between experimental and predicted ΔG . Development of a robust model is in progress.

3.5 | Mutational analysis

We have performed a mutational scanning for the epitope and paratope residues for each neutralizing antibody to identify mutants

that improve stability and binding affinity of the spike–antibody complex or vice versa. The analysis of changes in interaction energy upon point mutations showed that 85% of the residues in the epitope and paratope are important for binding as they reduce the binding affinity of the complex upon mutation. There are certain binding sites in antibodies, where mutation leads to increased binding affinity and stability of the paratope to the RBD. The mutation sites which improve affinity and stability for more than 10 amino acid mutations are given in Table 3. These paratope residues could be considered for engineering more potent antibodies against COVID-19.

The SARS-CoV-2 RBD is prone to many mutations that could escape the neutralization. We have identified several mutations in the RBD epitopes that decrease the binding affinity and stability of the antibodies (Table 3). The epitope residues which decrease the binding affinity and stability for more than 50% of mutations are F486, Y489, Q493, L455, and F456 (residues present in more than 50% of the antibodies). These positions have also been reported as having mutations emerging upon exposure (co-incubation) to mAbs.^{50,51} The mutation site F486 has shown immune escape from a mAb,

Variants of concern				
WHO label	Lineage + additional mutations	Country first detected (community)	Spike mutations of interest	Year and month first detected
Beta	B.1.351	South Africa	K417N, E484K, N501Y, D614G, A701V	September 2020
Gamma	P.1	Brazil	K417T, E484K, N501Y, D614G, H655Y	December 2020
Delta	B.1.617.2	India	L452R, T478K, D614G, P681R	December 2020
Variants of interest				
n/a	B.1.620	Unclear	S477N, E484K, D614G, P681H	February 2021
Mu	B.1.621	Colombia	R346K, E484K, N501Y, D614G, P681H	January 2021
Lambda	C.37	Peru	L452Q, F490S, D614G	December 2020

TABLE 4 Spike protein mutations in variants of concern and variants of interest

Note: The data are taken from European Center for Disease Prevention and Control (<https://www.ecdc.europa.eu/en/covid-19/variants-concern>). n/a: no WHO label has been assigned to this variant (accessed on October 6, 2021).

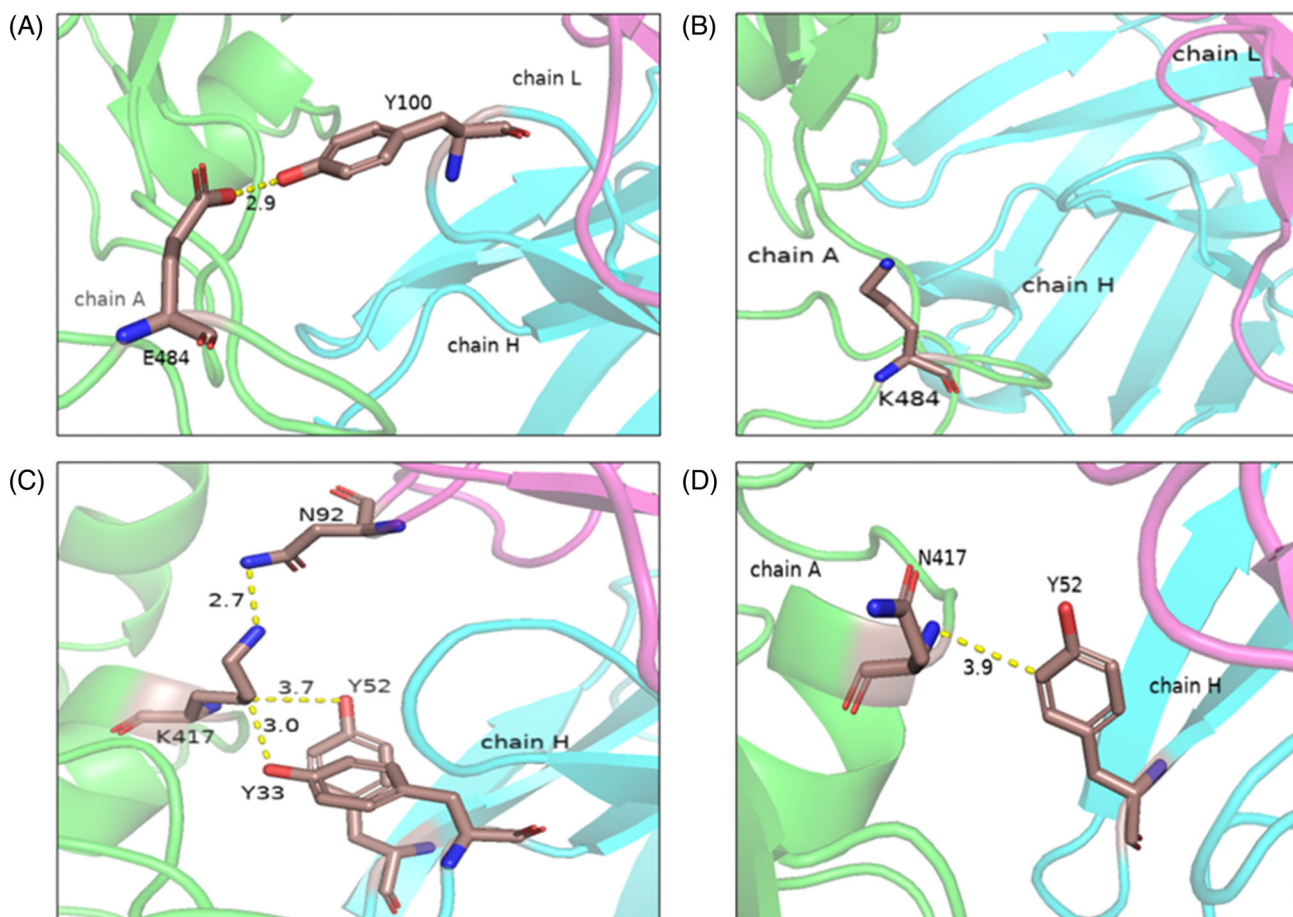


FIGURE 5 The interactions at the epitope positions and mutants (A) E484, (B) E484K, (C) K417, and (D) K417N in the spike-antibody complex (PDB: 7BZ5)

COV2-2832.⁹ Experimental studies have identified the key immune escape mutant residues as E484, S477, N439 with mutations E484Q, E484K, E484P, S477G, S477N, S477R, and N439K^{10,50} for a set of antibodies or convalescent sera, and we also observed a decrease in

binding affinity and stability at these sites upon mutation. The delta variant (B.1.617.2) showed reduced sensitivity to neutralization by antibodies.⁵² The amino acid changes in the RBD in this variant, K417N (present in delta plus, B.1.617.2.1), L452R, and T478K show a

decrease in binding affinity to antibodies in our study. We noticed that the residues K417, L452, and T478 within the epitopes are mutated frequently. Further, mutations at these sites reduced the binding free energy (Table 3). Among the Gamma variant (P.1) mutations,⁵³ K417T and N501Y are observed frequently in the RBD of the spike protein and these mutations decrease the binding affinity (Table 3). Our results are consistent with the observations that the variants of concern and variants of interest highlighted by WHO (Table 4) also decreased the binding free energy and stability.

We have further illustrated the results with two examples: (a) E484K and (b) K417N. E484 in the spike protein of the complex (PDB: 7BZ5) forms hydrogen bonds with Y100 of the antibody heavy chain (Figure 5A) and the mutation of E484K disrupt these interactions (Figure 5B) and reduced the binding affinity of 1.83 kcal/mol. On the other hand, K417 forms hydrogen bonds and cation- π interactions with Y33 and Y52 of the antibody heavy chain and hydrogen bonds with N92 of the antibody light chain (Figure 5C). The mutation K417N abolished the interactions with Y33 and N92 (Figure 5D) and decreased the binding affinity by 1.23 kcal/mol.

4 | CONCLUSIONS

Our study revealed that Tyr is the key residue involved in the binding of the spike and antibodies, which contributes to aromatic and hydrophobic interactions at the interface. YY pairs are the most preferred and energetically favored pairs present at the spike-antibody interface. The regression analysis to predict ΔG shows that interaction energy and atomic interactions play an important role. The mutational analysis of interface residues revealed that mutations in epitope sites F486, Y489, Q493, L455, and F456 decrease the binding affinity of spike-antibody complexes and could be considered as potential immune escape sites. In conclusion, the study will be helpful in understanding the interactions between the spike protein and the neutralizing antibodies in order to reveal the mechanisms of neutralization, which could help curb the infection and in developing better therapeutics against COVID-19. It could also provide an understanding of immune escape for emerging variants and on improving stability and binding affinity of neutralizing antibodies.

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AUTHOR CONTRIBUTIONS

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DATA AVAILABILITY STATEMENT

The data that support the findings of this study are available from the corresponding author upon reasonable request.

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