1 Structure, Dynamics and Free Energy Studies on the Effect of Spot Mutations on

2 SARS-CoV-2 Spike Protein Binding with ACE2 Receptor

3 Short title: Structure, Dynamics and Free Energy Studies on RBD Variants Binding with ACE2

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Abstract

The ongoing COVID-19 pandemic continues to infect people worldwide, and the virus continues to evolve in significant ways which can pose challenges to the efficiency of available vaccines and therapeutic drugs and cause future pandemic. Therefore, it is important to investigate the binding and interaction of ACE2 with different RBD variants. A comparative study using all-atom MD simulations was conducted on ACE2 binding with 8 different RBD variants, including N501Y, E484K, P479S, T478I, S477N, N439K, K417N and N501Y-E484K-K417N on RBD. Based on the RMSD, RMSF, and DSSP results, the overall binding of RBD variants with ACE2 is stable, and the secondary structures of RBD and ACE2 are consistent after the spot mutation. Besides that, a similar buried surface area, a consistent binding interface and a similar amount of hydrogen bonds formed between RBD with ACE2 although the exact residue pairs on the binding interface were modified. The change of binding free energy from spot mutation was predicted using the free energy perturbation (FEP) method. It is found that N501Y, N439K, and K417N can strengthen the binding of RBD with ACE2, while E484K and P479S weaken the binding, and S477N and T478I have negligible effect on the binding. Spot mutations modified the dynamic correlation of residues in RBD based on the dihedral angle covariance matrix calculation. Doing dynamic network analysis, a common intrinsic network community extending from the tail of RBD to central, then to the binding interface region was found, which could communicate the dynamics in the binding interface region to the tail thus to the other sections of S protein. The result can supply unique methodology and molecular insight on studying the molecular structure and dynamics of possible future pandemics and design novel drugs.

Introduction

COVID-19 pandemic which is the result of infection by SARS-Coronavirus-2 (CoV-2), had

claimed over 6 million lives as of June 2023, and continues impacting worldwide health [1]. CoV-

2 expresses S (Spike) protein [2,3], which is responsible for binding to the host cell receptor

followed by fusion of the viral and cellular membranes, respectively [4]. To engage a host cell

receptor, the receptor-binding domain (RBD) of the S protein undergoes hinge-like conformational

movements [5], thus mediates interaction with the receptor angiotensin-converting enzyme 2

(ACE2) [6,7,8,9] (the RBD and ACE2 binding structure is shown in Figure 1 (a)). The binding of

RBD with ACE2 is believed to be the critical initial event in the infection cascade. The interface

residues on RBD thus play key roles in the spike protein's binding with ACE2.

Coronaviruses are enveloped viruses that possess a positive-sense single-stranded RNA genome

26–32 kb in length [10], and have a remarkable mutation rate that allows them to evolve in a way

that perhaps by increasing their transmissibility [11,12, 13, 14] and/or their resistance to protective

immunity induced by previous infection or vaccines [15, 16, 17, 18, 19, 20]. Up to now, different

variants of concern (VOCs) have shown up, including Alpha, Beta, Gamma, Delta, and Omicron

[21,22,23,24,25,26,27,28] (Some mutations involved are shown in Table S1 in SI). The Alpha

variant was found in UK in September of 2020, while Beta variant was found in North Africa in

May of 2020. Gamma variant was found in Brazil [21,22,23,25]. The Omicron variant which

contains 11 mutants on RBD, is still the dominant cause of the resurgence of infection in many

countries currently [27]. Most of the mutations (around 58% of mutations) on RBD are located at

the receptor binding motif which is the binding interface with ACE2 [29].

N501Y on RBD, which appeared in Alpha, Beta, Gamma and Omicron variants, increased the

binding affinity of RBD with ACE2 [30, 31] and enhanced the viral resistance to neutralizing

antibodies [32]. S477N appeared in Omicron and E484K appeared in both Beta and Gamma

variants, were reported to enhance the binding affinity of S protein with ACE2 [33,34]. K417N

found in both Beta and Omicron variants was reported to decrease the binding affinity of RBD and

ACE2 [35], and was thought to cause immune escape [36]. The triple mutant of N501Y-E484K-

K417N appeared in Beta variant, has been shown to increase the infectivity of virus considering

the effect of each single mutant [35]. N439K, which was found in multiple regions in the world

and a quite common mutation, can enhance the binding of RBD with human ACE2 receptor while

evade antibody mediated immunity [37]. P479S, which also happened with high frequency at

most countries based on SARS-CoV-2 genome samples deposited at GISAID in early 2020 [38],

was predicted to make the virus less infective based on algebraic topology-based machine learning

modelling [29]. Besides that, based on the report of most frequently occurring mutations in RBD

in January 2021 (https://www.gisaid.org/hcov19-mutation-dashboard), T478I which appeared in

Alpha variant, however did not influence the binding of RBD and ACE2 [39]. Those mutants all

appeared on the RBD binding interface.

Besides those experimental work, in order to find out the mechanism of CoV-2 variants binding

with ACE2 and to design effective therapeutics to deal with CoV-2 variants, molecular dynamics

simulation methods have been applied to study RBD binding with ACE2 [33,40,41,34,42,

43,44,45,46,47]. Different simulation methods were used to predict the binding free energy change

($\Delta\Delta G$) [48, 49], including free energy perturbation (FEP) method [43,41], molecular mechanics

Poisson-Boltzmann surface area (MM-PBSA) method [42], molecular mechanics-generalized

Born surface area (MM/GBSA) method [45] and so on. Li et al. [49] and other researchers found

that FEP method can predict $\Delta\Delta G$ reasonably well [41,48, 50] and better than other methods [49].

Thus FEP method was applied to predict the spot mutation on RBD to the binding of RBD variants

with ACE2 in this work. Besides that, dynamic correlation was calculated to predict correlation

matrix and dynamic network in RBD in order to understand how the S spike mutations on RBD

binding interface can affect the binding of RBD with ACE2. Since spot mutations happened quite

often on CoV-2 RBM [29], a comparative study on multiple single mutations on RBD binding

with ACE2 is necessary. In total 7 RBD single-site mutations which happened naturally and can

potentially affect the binding and dynamics of RBD with ACE2 were studied including N501Y,

E484K, K417N, N439K, P479S, T478I, S477N, and one triple mutant N501Y-E484K-K417N.

Their effects on the binding of RBD with ACE2 were analyzed. Besides all-atom NAMD

simulations, FEP method was applied to predict $\Delta\Delta G$ of RBD with ACE2 coming from the spot

mutation. The simulation predictions helped to clarify the contradicting available experimental

results. Calculating the dihedral angle covariance matrices and doing dynamic network analysis,

an intrinsic network community was predicted. Our results revealed the molecular mechanism

that underlies the dynamics and energy change of some RBD variants binding with ACE2. Such

kind of information will be valuable for the development of further vaccines and neutralizing

antibodies against mutant forms of the SARS-CoV-2 virus.

Figure 1. The binding structure of ACE2 (in blue) with RBD (in green) (a), with the mutations

(shown in red sticks) on RBD labeled; and the comparison of the initial and final structures of

ACE2 bound with RBD-trimutant variant (b). In b), the initial structures of ACE2 and RBD are

shown in wheat, and the final structure of ACE2 is shown in raspberry and magenta for RBD. The

mutations of E484K, K417N, and N501Y are shown in blue sticks in both the initial and final

structures.

Materials and Methods

1). Based on the survey up to now, 7 naturally happened single RBD mutants were studied: E484K, N439K, P479S, T478I, N501Y, K417N, S477N, besides one RBD triple mutant which has E484K, N501Y and K417N. The selection of those mutations is based on different SARS-CoV-2 variants shown as listed in Table S1 in SI. The mutation sites on RBD and ACE2 binding interface are shown in Figure 1(a). The crystal structure of the RBD domain of the Spike protein is available in complex with ACE2 at 2.45 Å resolution in the PDB with ID 6M0J [51], which was used as the starting structure to build the RBD mutants bound with ACE2. VMD program [52] was applied to mutate residues on RBD to generate RBD mutants. The corresponding RBD and ACE2 structures in [51] were used in reference simulations of RBD and ACE2 in free forms respectively. The bound and free forms of simulations conducted in this project are listed in Table S2.

Simulation systems on RBD mutants bound with ACE2 were set up using VMD program and CHARMM36m forcefield [53]. Each system was solvated with enough amount of TIP3P water molecules with the minimum distance between protein and water box edge being at least 12 Å, and was neutralized with 0.15M NaCl solvent. After a brief energy minimization using the conjugate gradient and line search algorithm, 4 ps of dynamics was run at 50 K, and then the system was brought up to 310 K over an equilibration period of 1 ns running all-atom molecular dynamics simulations using NAMD program [54] and CHARMM36m forcefield [53]. The temperature was 310K and a desired pressure of 1 atm using standard thermo- and barostats. The sampling run was 100 ns with a time step of 1 fs for each system. The trajectory output frequency was 1 ps. To do comparisons with the binding of RBD mutants with ACE2, the ACE2 bound with RBD wildtype system, the ACE2 only in solvent system, and the RBD only in solvent system were

138 also set up and 100 ns simulation was also performed for them each. The simulation systems, set

139 up, the number of atoms and box size information are shown in Table S2.

140 To analyze the trajectories, the Root Mean Square Deviation (RMSD) and Fluctuations (RMSF)

141 of the proteins were calculated using the VMD program and an in-house analysis script based on

142 the coordinates of the backbone CA atoms after aligning the trajectories respectively, to the

143 original crystal structure of the RBD, ACE2, and to the initial complex structure of the RBD and

144 ACE2 from Lan et al.[51]. The buried surface area (BSA) for the complex was calculated in two

145 steps using the VMD program and a script using the Richards and Lee method with the water probe

146 size of 1.4 Å (Lee and Richards, 1971). First, the total solvent accessible surface area of the

147 complex ($ASA_{COMPLEX}$) was calculated based on the complex's trajectory. Second, the accessible

148 surface area of each protein in the complex (ASA_{RBD}, ASA_{ACE2}) was calculated for each protein

149 individually. Then, the buried surface area (BSA) is calculated using Equation (1):

151 $BSA = 0.5*(ASA_{RBD} + ASA_{ACE2} - ASA_{COMPLEX})$     (1)

153 The number of hydrogen bonds between the RBD and ACE2 were calculated using the VMD

154 program with the heavy atom distance cutoff of 3.0 Å and the angle cutoff of 20 degrees deviation

155 from H-bond linearity. The time a particular H-bond is formed over the course of the simulation

156 is monitored and is expressed as % occupancy. In order to find out the residues on the binding

157 interface, the closest distance between every residue atom (including hydrogen) between the RBD

158 and ACE2 was calculated and averaged over the trajectory run. The average distances between

159 each residue on RBD and on ACE2 are shaded by proximity on a red to white color-scale and were

160 used to build the distance maps.

2). Dihedral Angle Correlation, network community calculation. To quantitatively describe the

motional correlations of the backbone in RBD, we calculated the φ and ψ cross-correlation matrix

for all the residues except the first or last one on the sequence. φ describes the rotation of the

N−CA bond and involves the CO−N−CA−CO bonds. ψ describes the rotation of CA and the CO

bond and involves the N−CA−CO−N bonds. Both φ and ψ were calculated for residues at different

times using the wordom program [55,56]. Because both φ and ψ are angular variables, to avoid

the periodicity problem, the circular correlation coefficient, which is a T-linear dependence, was

calculated in this project following the method of Fisher [57,58] and Mardia and Jupp[59], and in

the same way as in our previous work [60,61]. The $\phi-\varphi$ cross-correlation coefficients (matrices)

were calculated of all combinations of dihedral angles. In order to simplify the data representation,

the $\phi-\phi$, $\phi-\varphi$, and $\varphi-\varphi$ correlation matrices were combined and the element with the largest

magnitude at residue pair position was always chosen to build the final matrix.

In the dynamic network community calculation, the network model was built by the NetworkView

plug-in of VMD [62] and the program Carma [63]. In the network, the nodes could be considered

as a single atom or a cluster of atoms. In this project, each CA atom in one amino acid was treated

as one node. The edge between two nodes was defined with the cutoff distance of 4.5Å for at least

75% of molecular dynamics simulated trajectory.

3). In order to find out the contribution of mutant residues on RBD to its binding with ACE2, FEP

method was applied in the same way as in [50]. The change of free energy of each mutation to the

binding was predicted, including E484K, N439K, N501Y, P479S, S477N, T478I, K417N on RBD

binding with ACE2.

In total, two kinds of systems were set up. One is the RBD in wildtype form in solvent (unbound

state). The other is the RBD mutant bound with ACE2 (bound state). For both kinds of systems,

mutation on one residue (E484K, N439K, K417N, T478I, P479S, S477N, N501Y) on RBD

(mutant systems) was conducted using VMD program [52]. After solvating the unbound and bound

systems with RBD in both wildtype and mutant forms with enough amount of water and

neutralizing them with 0.15M NaCl, NAMD all-atom molecular dynamics simulations were

performed to equilibrate both kinds of systems using an NVT ensemble after a brief energy

minimization. The dual-topology methodology [64,65,66], and the soft-core potentials [67,68]

were applied to overcome endpoint singularities [69] in the FEP simulation, following the same

method as in our previous work [50]. The free energy change in mutation for residue i in

thermodynamic cycle can be calculated using Equation (2):

$$\Delta G_{i,wildtype,bind} - \Delta G_{i,mutant,bind} = \Delta G_{i,unbound,alch} - \Delta G_{i,bound,alch} = \Delta\Delta G_i \quad (2)$$

Here, $\Delta\Delta G_i$ is the relative change of alchemical free energy of residue i during the binding of RBD

with ACE2 as the thermodynamic cycle shown in Figure S1. In each FEP simulation, 3 ns

simulations were performed in each window for both forward and backward directions and for

both the bound and free states after 0.5 ns simulations were conducted to equilibrate the system.

In this project, the interaction calculation includes 10 windows, ranging from a thermodynamic

coupling parameter λ values of zero to 1 (with a gap of 0.1) for a total of 20 simulations per

mutation. The time step was 1.0 fs.

Results

1). RMSD, RMSF, ΔRMSF, DSSP results for ACE2 and RBD variants

RMSD result. Based on all-atom NAMD simulations (100 ns each) on different systems, the

RMSDs of the complex, RBD, ACE2 were calculated after aligning the trajectories to the RBD

wildtype bound with ACE2, RBD wildtype, and ACE2 individually. RBD mutants, ACE2, and the

complexes are very stable during 100 ns simulations and are very consistent with those from ACE2

bound with RBD wildtype (as the RMSD vs simulation time result shown in Figure S2 in SI).

Calculating the average RMSD of the complex, the RBD, and the ACE2, the results are shown in

Table 1. Only ACE2&RBD-N501Y complex has a RMSD (2.09±0.2Å) lower than that of the

ACE2&RBD wildtype(2.21±0.4Å). Although RBD-N501Y has the highest RMSD (1.54±0.1Å),

ACE2 bound with this RBD variant has the lowest RMSD (1.84±0.2Å). That suggests that RBD-

N501Y can fit the binding pocket of ACE2 with the minimum structure change on ACE2 and the

maximum structure adaptation of RBD-N501Y. On the other hand, the complexes of ACE2 with

RBD-trimutant, RBD-T478I, and RBD-N439K showed the largest structural deviation from the

wildtype(3.05±0.4Å, 3.06±0.4Å, 3.34±0.5Å). When bound with RBD-N439K, ACE2 has the

largest structural deviation (2.60±0.3Å) comparing to the ACE2 wildtype (1.88±0.2Å). The

binding of RBD with ACE2 decreased the structural deviation of both RBD and ACE2. Overall,

the RMSD result suggests that spot mutation did not affect the binding structure of ACE2 with

RBD significantly.

Table 1. The average and standard deviations of RMSDs of ACE2&RBD complex, RBD wildtype

and variants, and ACE2, the number of hydrogen bonds (hbonds) formed, and the buried surface

area (BSA) between ACE2 and RBD. The standard deviations were calculated based on the

second-half of NAMD simulations.

Simulations	RMSD$_{Complex}$ (Å)	RMSD$_{RBD}$ (Å)	RMSD$_{ACE2}$ (Å)	Hbond	BSA (Å^2)
E484K&ACE2	2.40±0.3	1.22±0.1	1.92±0.1	2.63±1.7	864.3±57.4
N439K&ACE2	3.34±0.5	1.18±0.1	2.60±0.3	2.69±1.4	876.0±46.8
N501Y&ACE2	2.09±0.2	1.54±0.1	1.84±0.2	2.62±1.3	843.6±54.4
P479S&ACE2	2.74±0.7	1.21±0.1	2.11±0.2	3.23±1.5	861.0±62.0
K417N&ACE2	2.68±0.3	1.20±0.1	2.45±0.3	2.32±1.4	838.1±45.5
S477N&ACE2	2.89±0.4	1.15±0.1	2.50±0.2	3.03±1.4	908.6±41.4
T478I&ACE2	3.06±0.4	1.22±0.1	2.43±0.2	1.49±1.1	729.7±50.3
Tri-mutant&ACE2	3.05±0.4	1.32±0.2	2.10±0.2	2.01±1.2	835.5±46.5
RBD&ACE2 wt	2.21±0.4	1.23±0.1	1.88±0.2	2.64±1.4	863.3±44.4
RBD wt, free		1.52±0.2			
ACE2 free			2.62±0.4		

RMSF and ΔRMSF result. Calculating the Root Mean Squared Fluctuation (RMSF) of RBD

and ACE2 from different simulations, the results are shown in Figure 2(Left) and (Right).

Residues from 455 to 505 on RBD bound with ACE2 at the region of RES19 to RES81, which

are highlighted in light green for RBD on Figure 2(Left) and for ACE2 on Figure 2(Right).

Within the binding region, all the mutants have a smaller RMSF than the wildtype, especially

S477N, N439K, T478I and the tri-mutant. However, at other regions such as RES360 to

RES390, E484K showed a larger structure fluctuation.

Figure 2. Comparison of RMSF from RBD variants and wildtype (Left) and ACE2 (Right). The

variant results are shown in black solid line while the wildtype shown in red dashed line. The

binding regions on RBD and ACE2 are shaded in light green, and the secondary structures of RBD

and ACE2 are shown on the top of RMSF figures.

Mapping the change of RMSF (ΔRMSF=RMSF$_{mutant}$-RMSF$_{wt}$) on the secondary structure of RBD

and ACE2, the result for N501Y is shown in Figure 3(Left) and the tri-mutant bound with ACE2

in Figure 3(Right). Comparing to the wildtype, RBD-N501Y is slightly more rigid at RES501 site

and RES484 site, although other binding region has a similar rigidity to the wildtype form.

However, ACE2 is more rigid on the binding interface except the loop region which becomes more

flexible than the ACE2 when bound with RBD wildtype. Figure 3(Right) shows that part of the

binding interface on RBD becomes more rigid, part unchanged while the left region become more

flexible comparing to the RBD wildtype, while ACE2 also has part of the interface more rigid,

part unchanged, and other part more flexible comparing to the ACE2 bound with RBD wildtype.

Those are consistent with RMSF result in Figure 2. Similarly, the ΔRMSF of other RBD mutants

and ACE2 were calculated and mapped to the secondary structure of RBD and ACE2 (shown in

Figure S3 to S5). In summary, the binding interface of ACE2 molecule becomes more rigid when

bound with RBD variants except with S477N and N439K which have more flexible region than

rigid region on the binding interface. Comparing to the RBD wildtype, RBD variants only become

slightly more rigid except RBD-N439K variant which becomes mostly more flexible. Faraway

regions of ACE2 bound with RBD variants usually become more flexible (such as RBD-S477N,

RBD-N439K, and RBD-K417N, RBD-T478I), but not for the ACE2 bound with E484K, which

overall become more rigid. Those can agree with RMSF results shown in Figure 2, and the result

suggests that spot mutations on RBD can change the overall structure flexibility of RBD and ACE2

comparing to the wildtype. Such kind of change of flexibility should contribute to the adaptation

of RBD variants to fit the binding pocket of ACE2 which also can adjust its flexibility accordingly.

Figure 3. The ΔRMSF of RBD-N501Y and ACE2 (Left) from RBD wildtype to variant forms

mapped to the initial bound structure of ACE2&RBD-N501Y; and the ΔRMSF of RBD-

trimutant and ACE2 (Right) mapped to its initial bound structure of the complex. Colors range

from blue to white to red (with white representing no change in RMSF, blue representing an

RMSF decrease, and red representing an RMSF increase). The ΔRMSF is in the range from -0.5

(totally blue) to 0.5 (totally red).

DSSP result. In order to track the structure change of ACE2 and RBD contributed by the spot

mutation, the evolution of the secondary structure was predicted based on DSSP [70] calculation

over the simulation time using the wordom program. The DSSP program assigns the protein

secondary structure to one of the nine states defined, including three helix types (3_{10}, G; α, H; π,

I), two extended sheet types (antiparallel and parallel, E and X), two types of turns including the

helix turn (T) and β bridges (B), bend (S), and others (L). The secondary structure was calculated

for each residue at each time frame during the simulation time. Based on that, the average

proportion of helix (including G, H and I), sheet (including E), turn (including B and T) and other

(including X, S and L) structures on RBD and ACE2 over the simulation time were calculated,

and the results in comparison with available literature data [39] are shown in Figure 4 for RBD

and in Figure S6 in SI for ACE2.

Figure 4. Comparison of RBD DSSP with available literature data.

Overall, the proportions of RBD secondary structure elements are very consistent with available

experimental data [39], with around 13%-15% of helix, 30%-32% of sheet (including both parallel

and anti-parallel), and 13%-14% of turns, and 40%-43% of others. That implies the RBD structures

during NAMD simulations are reasonable. In Figure 4, comparing the DSSP of RBD mutants with

RBD wildtype from the experimental data (the last 6 columns which only have RBD in free state

available), it shows that the average percentage of other structure is higher in RBD mutants than

the RBD wildtype. That is consistent with the simulation prediction for RBD wildtype and variants

(the first 9 columns which are all in bound form). That suggests that spot mutation can increase

undefined structure (other structure) in RBD. Although DSSP of RBD in wildtype free state (wt,

free) has similar helix and turn ratios to the experimental data of RBD in wildtype free state (wt,

free, ex), its sheet ratio is higher than the experimental data by around 6% while its other structure

ratio is lower than the experimental data by around 6%. Such kind of discrepancy should come

from the different definitions of sheet from Anderson [70] and the experimental method. Because

of that, the average percentage of sheet structure from simulations are overall higher than

experimental data while the average percentage of other structures are overall lower than

experimental data.

Comparing results from simulation only or from experiments only, overall spot mutation only

changes the secondary structure distribution of RBD slightly in both bound and free states. It

almost does not change the secondary structure distribution of ACE2 as shown in Figure S6. So,

overall, the mutants and wildtype have similar secondary structure distribution and spot mutation

should only affect the secondary structure locally.

2). Hydrogen bonds formed on the binding interface

Forming intermolecular hydrogen bonds is one of the major driving forces for the binding of RBD

with ACE2. Calculating the number of hydrogen bonds formed between ACE2 and RBD mutants

over the simulation time (shown in Figure S7), overall RBD mutants formed similar amount of

hydrogen bonds with ACE2 to the RBD wildtype during the simulation time. Calculating the

average and standard deviation of the number of hydrogen bonds formed, the results are shown in

Table 1. RBD can form averagely 1.5 to 3.2 hydrogen bonds with ACE2. S477N and P479S formed

more hydrogen bonds with ACE2 than the wildtype, while T478I formed the least, and the tri-

mutant formed the second to the least amount of hydrogen bonds with ACE2. Tracking the residue

pairs on the binding interface, the results are shown in Table S3 in SI. There are 8 pairs of residues

forming hydrogen bonds on the RBD wildtype-ACE2 interface, including Gly502-Lys353,

Tyr505-Glu37, Asn487-Tyr83, Thr500-Tyr41, Lys417-Asp30, Gln493-Glu35, Tyr449-Asp38,

Gln498-Gln42. With the triple-mutation, three pairs (Lys417-Asp30, Tyr449-Asp38, Gln498-

Gln42) were broken and only the other five still formed hydrogen bonds on the tri-mutant&ACE2

interface. However, a new hydrogen bond formed between THR500-ASP355, and the interface

structures generated from ligplot program [71] in comparison with RBD wildtype are shown in

Figure 5. The details of hydrogen bonds formed between ACE2 and RBD wildtype and variants

are shown in Table S3. Overall, similar binding interface formed in RBD wt&ACE2 and RBD

variant&ACE2, although the exact residues forming hydrogen bonds changed and the number of

hydrogen bonds formed also changed by certain amount. Such kind of small modification on the

binding interface contributes to the different binding free energy of ACE2 and RBD which mainly

comes from hydrogen bonds formation and electrostatic and hydrophobic interactions [45], and

affects the infectivity of CoV-2 variants.

Figure 5. Hydrogen bonds formed on the ACE2&RBD interface for the wildtype (Top) and tri-

mutant (Bottom) generated from ligplot program. Residues formed hydrogen bonds are connected

using dashed green lines. Residues on RBD are labelled in light green with bonds shown in

magenta, while residues on ACE2 are labelled in blue with bonds shown in orange.

3). BSA and distance map result

BSA comparison. Calculating the buried surface area (BSA) between ACE2 and RBD variants

over the simulation time, the results in comparison with the RBD wildtype are shown in Figure 6.

RBD wildtype and variants can bind with ACE2 consistently during the simulation time.

Calculating the average and standard deviations of BSA based on the second half of the simulation

time, the results are shown in Table 1. Overall, the RBD variants and ACE2 form a similar BSA

as the RBD wildtype and ACE2. The RBD-T478I formed a slightly smaller BSA than other

mutants, while RBD-E484K, RBD-N439K and RBD-S477N form a larger BSA with ACE2 than

the RBD wildtype; the RBD-trimutant forms the smallest BSA among all the simulations

especially during 20 to 100 ns period in the simulations as shown in Figure 6. Based on the findings

from Chan et al. [43], a larger BSA suggests a stronger binding of RBD with ACE2, thus it is

expected that E484K, N439K and S477N variants should form a stronger binding with ACE2 than

other variants, while the tri-mutant should form the less stable binding with ACE2.

Figure 6. Comparison of the BSA between RBD variants and ACE2 with that of the ACE2&RBD

wildtype started from the crystal structure.

Distance map result comparison. Calculating the distance map between residues on ACE2 and

RBD, the results of RBD-E484K bound with ACE2 in comparison with RBD wildtype bound with

ACE2 are shown in Figure 7 as an example. The distance maps of other RBD variants bound with

ACE2 are shown in Figure S8 and S9 in SI. Those results show that different RBD variants bound

with ACE2 at the interface consistent with that of RBD wildtype and ACE2.

Even with the small modification on the residue pairs forming interactions on the binding interface,

a similar BSA and distance maps were formed for RBD variants with ACE2 comparing to RBD

wildtype. Thus it is not surprising to see that overall the binding structures of RBD mutants&ACE2

complexes are very consistent with the RBD wildtype complex which is the crystal structure of

RBD &ACE2 from Lan et al.[51], as the comparison of the initial and final structures for RBD-

trimutant&ACE2 shown in Figure 1 (b). The structures of RBD mutants bound with ACE2 showed

limited deviation from the original structure, which can agree with RMSD, RMSF, BSA and

hydrogen bonds results.

Figure 7. Distance map comparison of 1). RBD-E484K mutant bound with ACE2 and 2). RBD

wildtype bound with ACE2. The binding region (RES455 to 505) on RBD are highlighted in light

green. The color bars are shown on the right side of the distance maps, with a residue pair distance

being no larger than 4 Å shown in red, while a residue pair distance larger or equal to 14 Å shown

in white.

In summary, RBD mutants bind with ACE2 consistently and at a very similar interface. Because

the spot mutation did not change the binding interface of RBD and ACE2 significantly, it is

reasonable to start from the crystal structure of ACE2&RBD to do FEP simulation and predict the

spot mutation effect on $\Delta\Delta G$.

4). FEP result

Analyzing the FEP simulation results, $\Delta\Delta G$ was predicted for RBD mutants. The results in

comparison with Upadhyay [39] are shown in Figure 8. The comparison of simulation prediction

with three more available literatures [34,35] are shown in Table 2.

Figure 8. $\Delta\Delta G$ for RBD variants binding with ACE2 predicted from FEP (shown in blue bars) in

comparison with literature data from Upadhyay et al. (shown in orange bars). Reference $\Delta\Delta G$ was

calculated based on available data in the table as $\Delta\Delta G = \Delta G_{mutant} - \Delta G_{wt.}$

Table 2. $\Delta\Delta G$ predicted from FEP in comparison with available literature data.

Mutants	FEP prediction (kcal/mol)	Upadhyay et al. (kcal/mol)	Luan et al. (kcal/mol)	Laffeber (kcal/mol)	Barton et al. (kcal/mol)
N501Y	-3.80±0.17	-0.70±0.40	-0.81	1.3	1.43±0.04
N439K	-2.26±0.20				
K417N	-0.56±0.21	1.30±0.10		-0.8	-0.96±0.06
E484K	0.36±0.27	0.90±0.10		0.2	0.20±0.04
S477N	-0.08±0.09	-0.40±0.20			0.33±0.04

P479S	0.58±0.07				
T478I	0.19±0.08	0.3±0.20			

Based on Figure 8, N501Y, N439K, K417N contribute to the binding of RBD with ACE2, T478I

and S477N have a negligible effect, while neither E484K nor P479S contributes to the binding.

Upadhyay et al. [39] did ITC experiments on the binding of RBD mutants with ACE2. Calculating

$\Delta\Delta G$ using equation: $\Delta\Delta G=\Delta G_{mutant}-\Delta G_{wt}$ based on available data, $\Delta\Delta G$ data from Upadhyay are

shown in Figure 8 and Table 2 above. As can be seen from Upadhyay et al., N501Y can increase

the binding of RBD with ACE2, T478I has a negligible effect while E484K can make the binding

weaker. Our prediction of those three mutations can agree with Upadhyay et al, and the effect of

N501Y also agrees with Luan et al. [41] who conducted all-atom molecular dynamics simulations

and FEP on RBD-N501Y mutant binding with ACE2, and predicted the $\Delta\Delta G=-0.81$kcal/mol as

shown in Table 2.

However, our FEP predicted that S477N should have a negligible effect on RBD binding with

ACE2, while a stronger binding of RBD-S477N variant with ACE2 by Upadhyay et al. Although

our prediction deviated from Upadhyay et al, the effect of S477N from Upadhyay et al. was

opposite to the binding of RBD and ACE2 by Barton et al. Our FEP prediction on K417N is

opposite to Upadhyay et al's work, although the FEP prediction can agree with the experimental

work by Laffeber et al [72] and Barton et al. [73] as shown in Table 2. The $\Delta\Delta G$ result from

Laffeber et al. [72] as shown were calculated after converting experimental data K_D for wildtype

and mutants to ΔG using the relationship as: $\Delta G_{exp}=-RTlnK_D$ at T=298K.

N439K, which was found in multiple regions in the world and was a quite common mutation, can

enhance the binding of RBD with human ACE2 receptor while evade antibody mediated immunity

as found by Thomson et al. [37]. FEP predicted that N439K can decrease the binding free energy

of RBD with ACE2 thus should increase the binding affinity between them. Our prediction can

agree with Thomson et al.'s result [37].

Based on our FEP prediction, both N501Y and K417N can increase the binding affinity of RBD

and ACE2, but E484K has a negative effect on the binding. Because of that, the triple mutants of

K417A-E484A-N501A should contribute to the binding of RBD with ACE2 if assuming a linear

relationship between the effect of each mutation and the total effect. Because of that, it is expected

that ACE2 should bind with the triple mutant stronger than the wildtype. That agrees with the

infectivity of Beta variant. The FEP prediction on N501Y suggests that Alpha variant should be

more infectious than the original SARS-CoV-2. That also agrees with experimental and clinic

findings.

4). Dihedral angle correlation covariance result

In order to find out how spot mutations affect the dynamic correlation network of RBD, dihedral

angle ϕ, φ combined covariance matrix was calculated for RBD in both wildtype and variant forms,

and in both bound and free states. The dihedral angle covariance matrices revealed a set of highly

correlated residues in RBD-N501Y and RBD-trimutant, as presented in Figure 9 (Left) and (Right)

individually. The RBD wildtype bound with ACE2 covariance matrix is shown in the lower-right

triangle below the diagonal line while the covariance matrices of RBD-N501Y variant and tri-

mutant variant are shown in the upper-left triangle above the diagonal line in both Figure 9 (Left)

and (Right). RBD-N501Y variant lost the original correlation in RBD wildtype between RES501

and other residues as pointed out by the orange arrows, but other correlations in the range of

RES470 to RES483 became stronger in RBD-N501Y variant (pointed out by red arrows) which

were weak in the original RBD wildtype. Another region around RES370 (pointed out by red

arrows) which is also faraway from RES501 site, shows a strong correlation in RBD-N501Y

variant, although such kind of strong correlation was not present in the RBD wildtype. Overall, a

stronger correlation shows up in RBD-N501Y variant comparing to the RBD wildtype. Similarly,

RBD-trimutant (three mutations as N501Y, E484K, K417N) (Figure 9(Right)) has more discrete

correlations (pointed out by red arrows) which were not originally in the RBD wildtype, while the

original correlations between N501, E484 and K417 and other residues (pointed out by orange

arrows) were lost. RES501 region in RBD-tri-mutant is strongly correlated with other regions as

pointed out by red arrows.

Figure 9. Comparison of the averaged mixed dihedral angle covariance matrix for RBD-N501Y

(upper-left triangle above the diagonal line) with RBD wildtype in bound state (lower-right triangle)

(Left), and the averaged mixed dihedral angle covariance matrix for RBD-trimutation (upper-left

triangle above the diagonal line) in comparison with RBD in wildtype (Right). The color bars of

covariance map are shown on the right of the figures from blue to yellow corresponding to the

covariance coefficients in the range of -0.1 to 0.1. The secondary structure of RBD is shown on

the top of the covariance matrices, and the residue numbers of RBD are labelled on both the x and

y-axes.

Figure 10. Comparison of the averaged mixed dihedral angle covariance matrices for RBD-E484K

(Left) and RBD wildtype in free state (Right) (upper left triangle above the diagonal line) with

RBD wildtype in bound state (lower right). The color bars of covariance map are shown on the

right of the figures from blue to yellow corresponding to the covariance coefficients in the range

of -0.1 to 0.1. The secondary structure of RBD is shown on the top of the covariance matrices,

and the residue numbers of RBD are labelled on both the x and y-axes.

The correlation matrices for RBD-E484K variant and RBD wildtype in free form were analyzed

with results in comparison with RBD wildtype in bound state shown in Figure 10 (Left) and (Right),

and results for other RBD variants in Figure S10 to S12. It was interesting to see that RBD-E484K

has strong correlation in the head (RES334-374) and tail (RES520-526) regions which are faraway

from each other, comparing to the RBD wildtype. The RBD wildtype in free state has a stronger

correlation at its tail region but weaker correlation at its head region than the RBD wildtype in

bound state. RES372 region has strong correlation with other residues. Since the RBD in free state

simulation starts from the crystal structure of RBD (by removing the bound ACE2 in PDB ID of

6M0J) which is in open state, while binding of RBD with ligand can make the binding region more

rigid [47], that suggests that binding with ACE2 weakened the correlation of RBD in its tail region

which is the binding interface with ACE2.

Figure 11. Covariance coefficient between residue 373 (a)/residue 372 (b)/residue 480 (c)/ residue

501(d) and other residues mapped on the structure of RBD wildtype in bound state (a), RBD

wildtype in free state (b), and RBD-N501Y variant (c), and RBD-trimutant variant (d). Positive

covariance is shown in red while negative covariance in blue and the covariance equals to zero in

white. The range of covariance is in the range of -0.1 to 0.1. Residues have positive correlation

with the picked residue are labelled in red while in blue for negative correlation.

Similarly, other RBD variants have their correlation matrix changed comparing to the RBD

wildtype as results shown in Figure S10 to S12.

Since RES373 on RBD wildtype in bound form, RES372 on RBD wildtype in free form, RES480

on RBD-N501Y, and RES501 on RBD-trimutant have strong correlation with other residues as

shown in Figure 9 and Figure 10, those residues were picked as base for RBD in different

simulations correspondingly. Mapping the covariance coefficients of the base residue with other

residues on the b-factor of the secondary structure of RBD, the results for RBD wildtype in bound

form, RBD wildtype in free form, RBD-N501Y, RBD-trimutant are shown in Figure 11(a), (b),

(c), and (d) correspondingly (The secondary structures mapped with covariance coefficients for

other RBD variants not shown). In Figure 11, residues having a covariance coefficient higher than

0.05 were labelled, and in red for a positive correlation coefficient while in blue for a negative

correlation coefficient. Overall, RBD wildtype in free form, RBD-N501Y, RBD-trimutant all have

more correlations than the RBD wildtype in bound form. In RBD-N501Y shown in Figure 11(c),

RES480 has strong correlation with P479, G476, F456, S469, N440, I441, T500, Y501, P507,

G485, H519, C525, G526, L335, N334. Besides that, other residues labelled in red and blue also

have correlation with E484. Checking all the secondary structures mapped with covariance

coefficients (picking different base residues) for RBD in different forms, some common

correlations exist among residues, such as RES366/363, RES373/372, RES402/403, RES440/441,

RES443/444, RES465/464, RES475/476, RES485/484, RES516/517, RES519, RES521/522,

RES526/525 in different RBD covariance matrices. Those residues are at the head, binding

interface and the tail region, which are mostly α-helices or loops on RBD. Thus it suggests that

the helices and loops at the head, binding interface and tail regions have stronger dynamic

correlations than the β sheet region.

Figure 12. Community networks formed in the RBD-N501Y(Left), in wildtype bound form

(middle) and wildtype free form (Right) based on MD simulation and dynamical network

analysis. The RBD network structures are oriented in the same direction as the RBD shown in

Figure 1.

Because there are interactions between atoms inside a protein, the protein can be viewed as a

network. Treating the CA atom in each residue as a node, the interactions between nodes inside

the network can be analyzed. Multiple nodes with strong interactions can form one community,

which is a subnetwork that partitions the original network in the protein. The communication

communities can be calculated using VMD program as explained in the Materials and Methods

section. The dynamical network and communication communities in RBD wildtype and variants

were analyzed based on residue−residue interactions over time. It was found that usually 7 to 9

network communities formed in RBD in different forms and at different states. The results for

RBD-N501Y, RBD wildtype in bound form, RBD wildtype in free form are shown in Figure 12,

and in Figure S13 to S14 for other RBD variants. Interestingly, the largest community (shown in

blue) always extends from the tail region of RBD to the binding interface in the community maps.

The largest community includes the following residues in common: RES335, RES348, RES373,

RES428, RES437, RES471, RES490, RES493, RES495, RES496, RES516, RES519, RES521-

526. Those residues are at the N-term, central, binding region, and tail of RBD.

Such kind of extensive community showed up in all other RBDs as their network community maps

shown in Figure 12 and Figure S13 to S14. That suggests that there should be an intrinsic dynamic

network community inside RBD distributed from the tail region to the central, and to the binding

interface, which is consistent with the dihedral angle covariance result shown in Figure 11. Thus,

the dynamics at the RBD binding interface region should be transmitted to the tail of RBD thus to

the other sections of S protein. It is expected that disrupting of this intrinsic network can affect the

binding of RBD with ACE2 and with antibodies.

Discussion

Based on the findings from Chan et al. [43], a larger BSA suggests a stronger binding of RBD with

ACE2, thus it is expected that E484K, N439K and S477N variants should form a stronger binding

with ACE2 than other variants, while the RBD-trimutant should form a less stable binding with

ACE2 based on results shown in Figure 6. However, those cannot agree with the ΔΔG result

predicted from FEP as shown in Figure 8, which shows that N439K mutation should decrease the

binding free energy of RBD and ACE2, but not E484K. S477N mutation almost does not change

the binding free energy of RBD and ACE2. Thus, the size of BSA does not correlate to the binding

affinity of RBD and ACE2 based on our test in this project.

Up to now, different results on spot mutation effects on the binding affinity of RBD with ACE2

were reported. Tian et al. [74] found that N501Y mutation can increase the binding affinity of

RBD with ACE2 receptor using both experimental and simulation methods, thus a higher rate of

transmission of SARS-CoV-2 variant. However, they found that E484K did not contribute to the

binding of RBD with ACE2. That is different from Wang et al. [75], who found that E484K mutant

on RBD can increase the binding affinity of RBD with ACE2 receptor, while decrease the binding

affinity of RBD with antibodies. Upadhyay et al. [39] found E484K had a weaker binding affinity

to ACE2 than the wildtype, although N501Y had a stronger binding affinity to ACE2 and S477N

and T478I have a similar affinity to ACE2. Based on experimental test result, Wang et al. [76]

suggested that T478I and N501Y mutants can increase the infectivity of SARS-CoV-2, and can

potentially facilitate the transmission of SARS-CoV-2 virus to many other animal hosts. Therefore,

more attention should be paid to SARS-CoV-2 variants with these two mutations. In this project,

it has been found that spot-mutation on RBD do not change the binding interface between ACE2

and RBD comparing to the wildtype. However, each spot-mutation can make different

contributions to the binding free energy of RBD with ACE2. N501Y can increase the binding of

RBD with ACE2 significantly based on FEP prediction. That agrees with findings from some other

researchers as shown in Table 2. Such kind of discrepancy in experimental result emphasized the

importance of theoretical prediction and systematical comparative study on the SARS-CoV-2 spot-

mutation variants.

Based on the dihedral angle covariance matrix calculation, it was found that there are some

common correlations in RBD, no matter in variant or wildtype form, in bound or free state. Those

residues are mostly located in the largest network community predicted by dynamic network

analysis with results shown in Figure 12 and Figure S13 to S14. Since the community is intrinsic

in RBD, and it extends from tail region of RBD to the binding interface region, it is expected that

mutation on those residues can affect the binding of RBD with ACE2 and with antibodies.

Since the spike protein of SARS-CoV-2 is a trimer and each monomer has more than 1000 residues

[77], more work should be done in order to understand how the spot mutations on S protein affect

its structure, dynamics and function thus to supply insight on designing novel drugs against future

pandemic.

Conclusions

In this project, all-atom molecular dynamics simulations using NAMD program were conducted

on seven RBD single mutants and one RBD triple mutant binding with ACE2. It was found that

spot mutation on RBD does not affect the binding interface of RBD with ACE2 significantly.

Similar number of hydrogen bonds still formed on the binding interface between RBD and ACE2

after the spot mutation, although some of the exact residues forming hydrogen bonds changed. The

distance map result shows that RBD mutant and ACE2 form the binding interface consistent with

the RBD wildtype, and they also form a similar buried surface area on the interface. Thus, the

structures of RBD mutants bound with ACE2 are mostly consistent with RBD wildtype with ACE2.

Calculating $\Delta\Delta G$ of spot mutation using FEP method, it was found that N501Y, N439K, K417N

can improve the binding of RBD with ACE2; E484K and P479S could not increase the binding of

RBD with ACE2; S477N and T478I do not affect the binding of RBD with ACE2. The spot

mutation changed the dynamic correlation network of residues on RBD based on the dihedral

correlation calculation, but an intrinsic network community exists in RBD which extends from the

tail region to the central region, then to the binding interface region. The result in this project can

supply the methodology and molecular insight on studying the molecular structure and dynamics

of possible future pandemics and design novel drugs.

Acknowledgements

This work was supported by National Science Foundation (NSF) Grant NO. 2200138. The

simulations were performed using supercomputer time from XSEDE via an award to Zhang L.

(MCB160041), and with some trajectory analysis partially using the high performance computers

at UTC, and one simulation performed on Unity computers of URI.

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wildtype

Trimutant