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

SARS-CoV-2 Spike Protein Binding with ACE2 Receptor

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 the binding of RBD variants with ACE2 is stable, and the secondary structure 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 mutants of N501Y-E484K-

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

the effect of each single mutants [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 in 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 [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 and ACE2 in order to understand how the S spike mutants 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 mutants 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 because of 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 tri-mutant 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

mutants 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 triple mutants of RBD 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 ACE2 and RBD 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.

Different simulation systems were set up using VMD program and CHARMM36m forcefield [53]. The 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 using NAMD program[54], all-atom molecular dynamics simulation was performed on each system using NAMD program and CHARMM36 forcefield [55]. 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. The trajectory output frequency was 1 ps. To do comparisons on the binding of RBD mutant with ACE2, the RBD bound with ACE2 wildtype systems were also set up and simulations were performed for 100 ns for RBD mutant & ACE2.

The simulation systems, set up, the number of atoms and box size information are shown in Table

S2.

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

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

the coordinates of the backbone Ca atoms after aligning the trajectories respectively, to the original

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

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

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

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

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

of each protein in the complex (ASA_{rbd}, ASA_{ACE2}) was calculated for each protein individually.

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

$$BSA = 0.5*(ASA_{rbd} + ASA_{ACE2} - ASA_{complex}) \qquad (1)$$

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

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

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

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

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

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

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

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−Cα

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

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

using the wordom program [56,57]. 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 [58,59] and Mardia and Jupp[60], and in

the same way as in our previous work [61,62]. The ϕ−φ cross-correlation coefficients (matrices)

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

the ϕ−ϕ, ϕ−φ, and φ−φ 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 [63] and the program Carma [64]. In the network, the nodes could be considered

as a single atom or a cluster of atoms. In this project, each Cα 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 has been 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 systems

with enough amount of water and neutralizing them with 0.15M NaCl, and solvating the bound

system with the same amount of water and enough amount of ions on both the wildtype and mutant

systems, 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 [65,66,67], and the soft-core potentials [68,69] were applied to overcome endpoint

singularities [70] 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 on different RBD variants bound with ACE2,

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 tri-mutant, 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). 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

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 free 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 results in RMSF result 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 wildtype to variant forms of RBD

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 RMSF

decrease, and red representing 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 [71] 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 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 [71] 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. 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 residues 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 and

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 in the tri-mutant&ACE2 interface. However, a new hydrogen bond

formed between THR500-ASP355, and the interface structure generated from ligplot program [72]

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 other RBD variants bound with ACE2 distance

maps 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 wildtype 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 consistent 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 FEP simulation results, ΔΔG 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. ΔΔG 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 ΔΔG was

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

Table 2. ΔΔG 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 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

ΔΔG using equation: $\Delta\Delta G = \Delta G_{mutant} - \Delta G_{wt}$ based on available data, ΔΔG 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 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 ΔΔG=-0.81kcal/mol as

shown in Table 2.

However, FEP predicted that S477N should have 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. The 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 [73] and Barton et al. [74] as shown in Table 2. The ΔΔG result from Laffeber et al.

[73] as shown were calculated after converting experimental data K_D for wildtype and mutants to

ΔG using the relationship as: $\Delta G_{exp} = -RT\ln K_D$ 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 for RBD was calculated in both the 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-tri-mutant, 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) in comparison of RBD in wildtype (lower-right triangle) in bound state

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

above the diagonal line) in comparison of 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 coefficient in the

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

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

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) in comparison 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 coefficient in the range of -0.1 to 0.1. The

secondary structure of RBD is shown on the top the covariance matrices, and the residue numbers of RBD

are labelled on both the x and y-axis.

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 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) and

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

(a), in the wildtype and free state (b), and in the N501Y variant (c), and in tri-mutant 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 bound form, RES372 on RBD wildtype 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 in wildtype bound form, in

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

individually (The secondary structures mapped with covariance coefficients for other RBD

variants are 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 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 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. That may suggest that mutating those residues will modify the dynamics network of RBD,

thus affect its binding with ACE2.

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 inside the protein, the protein can be viewed as a network. Analyzing

the interaction inside the network, communication communities can be calculated. Using the plug-

in in the VMD 1.9.3 edition, the dynamical network in RBD wildtype and variants was analyzed

based on residue−residue interactions over time. The Cα atom in each residue was treated as one

node. Multiple nodes with strong interactions can form one community. 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 bound form, RBD wildtype 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. Thus, the dynamics in 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 tri-mutant should form a less stable binding with ACE2

based on results shown in Figure 6. However, those cannot agree with the $\Delta\Delta G$ result predicted

from FEP as shown in Figure 8, which shows that N439K should decrease the binding free energy

of RBD and ACE2, but not E484K. S477N 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. [75] 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, it was found that E484K did not contribute to the

binding of RBD with ACE2. That is different from Wang et al. [76], who found that E484K mutant

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

binding affinity of RBD with antibodies. Upadhyay et al. [39] found that mutants differ in their

expression levels. Of the eight best-expressed mutants, two (N501Y and K417T/E484K/N501Y)

showed stronger affinity to ACE2 compared with the wildtype, whereas four (Y453F, S477N,

T478I, and S494P) had similar affinity and two (K417N and E484K) had weaker affinity than the

wildtype. Compared with the wildtype, four mutants (K417N, Y453F, N501Y, and

K417T/E484K/N501Y) had weaker affinity for the CC12.1 antibody, whereas two (S477N and

S494P) had similar affinity, and two (T478I and E484K) had stronger affinity than the wildtype.

Mutants also differ in their thermal stability, with the two least stable mutants showing reduced

expression. Taken together, these results indicate that multiple factors contribute toward the

natural selection of variants, and all these factors need to be considered to understand the evolution

of the virus. In addition, since not all variants can escape a given neutralizing antibody, antibodies

to treat new variants can be chosen based on the specific mutations in that variant. Wang et al. [77]

indicated that T478I and N501Y substitutions were two S mutations important for receptor

adaption of SARS-CoV-2, potentially contributing to the spillover of the 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 some

other researchers as shown in Table 2. Such kind of discrepancy 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 results 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

[78], 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 mutations on RBD do not affect the binding interface of RBD with ACE2 significantly.

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

ACE2 after 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 very consistent with RBD wildtype with

ACE2. Calculating ΔΔG 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.

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References:

1. WHO (2021) Coronavirus (COVID-19) dashboard Accessed March 22, 2023. Webpage: https://www.who.int/emergencies/diseases/novel-coronavirus-2019.

2. Siu YL, Teoh KT, Lo J, Chan CM, Kien F, Escriou N, Tsao SW, Nicholls JM, Altmeyer R, Peiris J SM, Bruzzone R, Nal B. The M, E, and N structural proteins of the severe acute respiratory syndrome

coronavirus are required for efficient assembly, trafficking, and release of virus-like particles. J Virol. 2008; 82:11318-11330.

3. Yoshimoto FK. The Proteins of Severe Acute Respiratory Syndrome Coronavirus-2 (SARS CoV-2 or n-COV19), the Cause of COVID-19. Protein J. 2020; 39:198-216.

4. Walls AC, Tortorici MA, Bosch B-J, Frenz B, Rottier PJM, DiMaio F, Rey FA, Veesler D. Cryo-electron microscopy structure of a coronavirus spike glycoprotein trimer. Nature. 2016; 531:114-117.

5. Wrapp D, Wang N, Corbett KS, Goldsmith JA, Hsieh C-L, Abiona O, Graham BS, McLellan JS. Cryo-EM structure of the 2019-nCoV spike in the prefusion conformation. Science. 2020; 367:1260-1263.

6. Walls AC, Park Y-J, Tortorici MA, Wall A, McGuire AT, Veesler D. Structure, Function, and Antigenicity of the SARS-CoV-2 Spike Glycoprotein. Cell. 2020; 181: 281-292.e286.

7. Yan R, Zhang Y, Li Y, Xia L, Guo Y, Zhou Q. Structural basis for the recognition of SARS-CoV-2 by full-length human ACE2. Science. 2020; 367: 1444-1448.

8. Lan J, Ge J, Yu J, Shan S, Zhou H, Fan S, Zhang Q, Shi X, Wang Q, Zhang L, Wang X. Structure of the SARS-CoV-2 spike receptor-binding domain bound to the ACE2 receptor. Nature. 2020; 581: 215-220.

9. McCallum M, Walls AC, Bowen JE, Corti D, Veesler D. Structure-guided covalent stabilization of coronavirus spike glycoprotein trimers in the closed conformation. Nat Struct Mol Biol. 2020; 27: 942–949.

10. Masters PS. The molecular biology of coronaviruses. Adv. Virus Res. 2006; 66: 193–292.

11. Davies NG, Abbott S, Barnard RC, Jarvis CI, Kucharski AJ, Munday JD, Pearson CAB, Russell TW, Tully DC, Washburne AD, Wenseleers T, Gimma A, Waites W, Wong KLM, van Zandvoort K, Silverman JD, CMMID COVID-19 Working Group, COVID-19 Genomics UK (COG-UK) Consortium, Diaz-Ordaz K, Keogh R, Eggo RM, Funk S, Jit M, Atkins KE, Edmunds WJ. Estimated transmissibility and impact of SARS-COV-2 lineage b.1.1.7 in England. Science. 2021; 372: abg3055.

12. Korber B, Fischer WM, Gnanakaran S, Yoon H, Theiler J, Abfalterer W, Hengartner N, Giorgi EE, Bhattacharya T, Foley B, Hastie KM, Parker MD, Partridge DG, Evans CM, Freeman TM, de Silva TI,

13. Volz E, Hill V, McCrone JT, Price A, Jorgensen D, O'Toole Á, Southgate J, Johnson R, Jackson B, Nascimento FF, Rey SM, Nicholls SM, Colquhoun RM, da Silva Filipe A, Shepherd J, Pascall DJ, Shah R, Jesudason N, Li K, Jarrett R, Pacchiarini N, Bull M, Geidelberg L, Siveroni I, COG-UK Consortium, Goodfellow I, Loman NJ, Pybus OG, Robertson DL, Thomson EC, Rambaut A, Connor TR. Evaluating the effects of SARS-COV-2 spike mutation D614G on transmissibility and pathogenicity. Cell. 2021b; 184: 64–75.

14. Washington NL, Gangavarapu K, Zeller M, Bolze A, Cirulli ET, Schiabor Barrett KM, Larsen BB, Anderson C, White S, Cassens T, Jacobs S, Levan G, Nguyen J, Ramirez JM, Rivera-Garcia C, Sandoval E, Wang X, Wong D, Spencer E, Robles-Sikisaka R, Kurzban E, Hughes LD, Deng X, Wang C, Servellita V, Valentine H, De Hoff P, Seaver P, Sathe S, Gietzen K, Sickler B, Antico J, Hoon K, Liu J, Harding A, Bakhtar O, Basler T, Austin B, MacCannell D, Isaksson M, Febbo PG, Becker D, Laurent M, McDonald E, Yeo GW, Knight R, Laurent LC, de Feo E, Worobey M, Chiu CY, Suchard MA, Lu JT, Lee W, Andersen KG. Emergence and rapid transmission of SARS-COV-2 b.1.1.7 in the United States. Cell. 2021; 184: 2587–2594.

15. Darby AC, Hiscox JA. Covid-19: Variants and vaccination. BMJ. 2021; 372: n771.

16. Dejnirattisai W, Zhou D, Supasa P, Liu C, Mentzer AJ, Ginn HM, Zhao Y, Duyvesteyn HME, Tuekprakhon A, Nutalai R, Wang B, López-Camacho C, Slon-Campos J, Walter TS, Skelly D, Costa Clemens SA, Naveca FG, Nascimento V, Nascimento F, Fernandes da Costa C, Resende PC, Pauvolid-Correa A, Siqueira MM, Dold C, Levin R, Dong T, Pollard AJ, Knight JC, Crook D, Lambe T, Clutterbuck E, Bibi S, Flaxman A, Bittaye M, Belij-Rammerstorfer S, Gilbert SC, Carroll MW, Klenerman P, Barnes E, Dunachie SJ, Paterson NG, Williams MA, Hall DR, Hulswit RJG, Bowden TA, Fry EE, Mongkolsapaya

Sheffield COVID-19 Genomics Group, McDanal C, Perez LG, Tang H, Moon-Walker A, Whelan SP, LaBranche CC, Saphire EO, Montefiori DC. Tracking changes in sars-cov-2 spike: Evidence that d614g increases infectivity of the COVID-19 virus. Cell. 2020; 182: 812–827.

J, Ren J, Stuart DI, Screaton GR. Antibody evasion by the p.1 strain of SARS-COV-2. Cell. 2021; 184: 2939–2954.

17. Garcia-Beltran WF, Lam EC, KSt D, Nitido AD, Garcia ZH, Hauser BM, Feldman J, Pavlovic MN, Gregory DJ, Poznansky MC, Sigal A, Schmidt AG, Iafrate AJ, Naranbhai V, Balazs AB. Multiple SARS-CoV-2 variants escape neutralization by vaccine-induced humoral immunity. Cell. 2021; 184: 2372–2383.

18. Madhi SA, Izu A, Pollard AJ. ChAdOx1 nCoV-19 Vaccine Efficacy against the B.1.351 Variant. The New England Journal of Medicine. 2021a; 385: 571–572.

19. Madhi SA, Baillie V, Cutland CL, Voysey M, Koen AL, Fairlie L, Padayachee SD, Dheda K, Barnabas SL, Bhorat QE, Briner C, Kwatra G, Ahmed K, Aley P, Bhikha S, Bhiman JN, Bhorat AE, du PJ, Esmail A, Groenewald M, Horne E, Hwa SH, Jose A, Lambe T, Laubscher M, Malahleha M, Masenya M, Masilela M, McKenzie S, Molapo K, Moultrie A, Oelofse S, Patel F, Pillay S, Rhead S, Rodel H, Rossouw L, Taoushanis C, Tegally H, Thombrayil A, van ES, Wibmer CK, Durham NM, Kelly EJ, Villafana TL, Gilbert S, Pollard AJ, de OT, Moore PL, Sigal A, Izu A. Safety and Efficacy of the Chadox1 NCOV-19 (AZD1222) Covid-19 Vaccine against the b.1.351 Variant in South Africa. medRxiv. 2021b.

20. Mahase E. Covid-19: Where are we on vaccines and variants. BMJ. 2021; 372: n597.

21. Davies N. G., Abbott S., Barnard R.C., Jarvis C.I., Kucharski A.J., Munday J.D., Pearson C.A.B, Russell T.W., Tully D.C., Washburne A.D., et al. Estimated transmissibility and impact of SARS-CoV-2 lineage B.1.1.7 in England. Science. 2021; 372: eabg3055. doi: 10.1126/science.abg3055.

22. Tegally H., Wilkinson E., Giovanetti M., Iranzadeh A., Fonseca V., Giandhari J., Doolabh D., Pillay S., San E.J., Msomi N., et al. Detection of a SARS-CoV-2 variant of concern in South Africa. Nature. 2021; 592: 438–443. doi: 10.1038/s41586-021-03402-9.

23. Faria N.R., Mellan T.A., Whittaker C., Claro I. M., Candido D.d.S., Mishra S., Crispim M.A.E., Sales F.C.S., Hawryluk I., McCrone J.T., et al. Genomics and epidemiology of the P.1 SARS-CoV-2 lineage in Manaus,Brazil. Science. 2021; 372: 815–821. doi: 10.1126/science.abh2644.

24. Fontanet A., Autran B., Lina B., Kieny M.P., Karim S.S.A, Sridhar D.J.T.L. SARS-CoV-2 variants and ending the COVID-19 pandemic. Lancet. 2021; 397: 952–954. doi: 10.1016/S0140-6736(21)00370-6.

25. Kannan S.R., Spratt A.N., Cohen A.R., Naqvi S.H., Chand H.S., Quinn T.P., Lorson C.L., Byrareddy S.N., Singh K. Evolutionary analysis of the Delta and Delta Plus variants of the SARS-CoV-2 viruses. J. Autoimmun. 2021; 124: 102715. doi: 10.1016/j.jaut.2021.102715.

26. Torjesen I. COVID-19: Omicron may be more transmissible than other variants and partly resistant to existing vaccines, scientists fear. BMJ. 2021; 375: n2943. doi: 10.1136/bmj.n2943.

27. Gao S.J., Guo H., Luo G. Omicron variant (B.1.1.529) of SARS-CoV-2, a global urgent public health alert! J. Med. Virol. 2021; 94: 1255–1256. doi: 10.1002/jmv.27491.

28. Han P., Li L., Liu S., Wang Q., Zhang D., Xu Z., Han P., Li X., Peng Q., Su C., et al. Receptor binding and complex structures of human ACE2 to spike RBD from omicron and delta SARS-CoV-2. Cell. 2022; 185: 630–640.e10. doi: 10.1016/j.cell.2022.01.001.

29. Chen J, Wang R, Wang M, Wei GW. Mutations Strengthened SARS-CoV-2 Infectivity. J Mol Biol. 2020; 432(19): 5212-5226. doi: 10.1016/j.jmb.2020.07.009.

30. Starr TN, Greaney AJ, Hilton SK, Ellis D, Crawford KHD, Dingens AS, Navarro MJ, Bowen JE, Tortorici MA, Walls AC, King NP, Veesler D, Bloom JD. Deep mutational scanning of SARS-COV-2 receptor binding domain reveals constraints on folding and ace2 binding. Cell. 2020; 182: 1295–1310.

31. Supasa P, Zhou D, Dejnirattisai W, Liu C, Mentzer AJ, Ginn HM, Zhao Y, Duyvesteyn HME, Nutalai R, Tuekprakhon A, Wang B, Paesen GC, Slon-Campos J, López-Camacho C, Hallis B, Coombes N, Bewley KR, Charlton S, Walter TS, Barnes E, Dunachie SJ, Skelly D, Lumley SF, Baker N, Shaik I, Humphries HE, Godwin K, Gent N, Sienkiewicz A, Dold C, Levin R, Dong T, Pollard AJ, Knight JC, Klenerman P, Crook D, Lambe T, Clutterbuck E, Bibi S, Flaxman A, Bittaye M, Belij-Rammerstorfer S, Gilbert S, Hall DR, Williams MA, Paterson NG, James W, Carroll MW, Fry EE, Mongkolsapaya J, Ren J, Stuart DI, Screaton GR. Reduced neutralization of SARS-COV-2 b.1.1.7 variant by convalescent and vaccine Sera. Cell. 2021; 184: 2201–2211.

32. Xie X., Liu Y., Liu J., et al. Neutralization of SARS-CoV-2 spike 69/70 deletion, E484K and N501Y variants by BNT162b2 vaccine-elicited sera. Nat Med. 2021; 27: 620–621. https://doi.org/10.1038/s41591-021-01270-4.

33. Wang WB, Liang Y, Jin YQ, Zhang J, Su JG, Li QM. E484K mutation in SARS-CoV-2 RBD enhances binding affinity with hACE2 but reduces interactions with neutralizing antibodies and nanobodies: Binding free energy calculation studies. J Mol Graph Model. 2021; 109: 108035. doi: 10.1016/j.jmgm.2021.108035. Epub 2021 Sep 17. PMID: 34562851; PMCID: PMC8447841.

34. Nelson G, Buzko O, Spilman P, Niazi K, Rabizadeh S, Soon-Shiong P. Molecular dynamic simulation reveals E484K mutation enhances spike RBD-ACE2 affinity and the combination of E484K, K417N and N501Y mutations (501Y.V2 variant) induces conformational change greater than N501Y mutant alone, potentially resulting in an escape mutant. bioRxiv 2021.01.13.426558; doi: https://doi.org/10.1101/2021.01.13.426558.

35. Barton MI, MacGowan SA, Kutuzov MA, Dushek O, Barton GJ, van der Merwe PA. Effects of common mutations in the SARS-CoV-2 Spike RBD and its ligand, the human ACE2 receptor on binding affinity and kinetics. Elife. 2021;10: e70658. doi: 10.7554/eLife.70658. PMID: 34435953; PMCID: PMC8480977.

36. Wang Z., Schmidt F., Weisblum Y., et al. mRNA vaccine-elicited antibodies to SARS-CoV-2 and circulating variants. Nature. 2021; 592: 616–622.

37. Thomson EC, Rosen LE, Shepherd JG, Spreafico R, da Silva Filipe A, Wojcechowskyj JA, Davis C, Piccoli L, Pascall DJ, Dillen J, Lytras S, Czudnochowski N, Shah R, Meury M, Jesudason N, De Marco A, Li K, Bassi J, O'Toole A, Pinto D, Colquhoun RM, Culap K, Jackson B, Zatta F, Rambaut A, Jaconi S, Sreenu VB, Nix J, Zhang I, Jarrett RF, Glass WG, Beltramello M, Nomikou K, Pizzuto M, Tong L, Cameroni E, Croll TI, Johnson N, Di Iulio J, Wickenhagen A, Ceschi A, Harbison AM, Mair D, Ferrari P, Smollett K, Sallusto F, Carmichael S, Garzoni C, Nichols J, Galli M, Hughes J, Riva A, Ho A, Schiuma M, Semple MG, Openshaw PJM, Fadda E, Baillie JK, Chodera JD; ISARIC4C Investigators; COVID-19 Genomics UK (COG-UK) Consortium; Rihn SJ, Lycett SJ, Virgin HW, Telenti A, Corti D, Robertson DL,

Snell G. Circulating SARS-CoV-2 spike N439K variants maintain fitness while evading antibody-mediated immunity. Cell. 2021; 184(5): 1171-1187.e20. doi: 10.1016/j.cell.2021.01.037.

38. Shu Y., McCauley J. GISAID: Globalinitiative on sharing all influenza data–from vision to reality. Eurosurveillance. 2017; 22(13).

39. Upadhyay V, Lucas A, Panja S, Miyauchi R, Mallela KMG. Receptor binding, immune escape, and protein stability direct the natural selection of SARS-CoV-2 variants. J Biol Chem. 2021; 297(4): 101208. doi: 10.1016/j.jbc.2021.101208. Epub 2021 Sep 17. PMID: 34543625; PMCID: PMC8445900.

40. Alaofi A. L., Shahid M. Mutations of SARS-CoV-2 RBD May Alter Its Molecular Structure to Improve Its Infection Efficiency. Biomolecules. 2021; 11: 1273. doi:10.3390/biom11091273.

41. Luan B, Wang H, Huynh T. Enhanced binding of the N501Y-mutated SARS-CoV-2 spike protein to the human ACE2 receptor: insights from molecular dynamics simulations. FEBS Lett. 2021; 595(10):1454-1461. doi: 10.1002/1873-3468.14076. Epub 2021 Apr 3. PMID: 33728680; PMCID: PMC8250610.

42. Kodchakorn K., Kongtawelert P. Molecular dynamics study on the strengthening behavior of Delta and Omicron SARS-CoV-2 spike RBD improved receptor-binding affinity. PloS one. 2022; 17(11): e0277745. doi:10.1371/journal.pone.0277745.

43. Chan KC, Song Y, Xu Z, Shang C, Zhou R. SARS-CoV-2 Delta Variant: Interplay between Individual Mutations and Their Allosteric Synergy. Biomolecules. 2022; 12(12): 1742. https://doi.org/10.3390/biom12121742.

44. Kim S, Liu Y, Ziarnik M, Seo S, Cao Y, Zhang XF, Im W. Binding of human ACE2 and RBD of Omicron enhanced by unique interaction patterns among SARS-CoV-2 variants of concern. Journal of Computational Chemistry. 2023; 44: 594-601.

45. Jawad B, Adhikari P, Podgornik R, Ching W. Key Interacting Residues between RBD of SARS-CoV-2 and ACE2 Receptor: Combination of Molecular Dynamics Simulation and Density Functional Calculation. J. Chem. Inf. Model. 2021; 61(9): 4425–4441.

46. Dutta S, Panthi B, Chandra A. All-Atom Simulations of Human ACE2-Spike Protein RBD Complexes for SARS-CoV-2 and Some of its Variants: Nature of Interactions and Free Energy Diagrams for Dissociation of the Protein Complexes. The journal of physical chemistry. B. 2022; 126(29): 5375–5389. https://doi.org/10.1021/acs.jpcb.2c00833.

47. Zhang L, Ghosh S K, Basavarajappa S C, Muller-Greven J, Penfield J, Brewer A, Ramakrishnan P, Buck M, Weinberg A. HBD-2 binds SARS-CoV-2 RBD and blocks viral entry: Strategy to combat COVID-19. iScience. 2022; 25: 103856.

48. Luan B, Huynh T. In Silico Antibody Mutagenesis for Optimizing Its Binding to Spike Protein of Severe Acute Respiratory Syndrome Coronavirus 2. J. Phys. Chem. Lett. 2020; 11: 9781-9787. doi:10.1021/acs.jpclett.0c02706.

49. Li Z, Zhang J Z H. Mutational Effect of Some Major COVID-19 Variants on Binding of the S Protein to ACE2. Biomolecules. 2022; 12: 572. doi:10.3390/biom12040572.

50. Brewer A, Zhang L. Binding Free Energy Calculation on Human Beta Defensin 3 on Negatively Charged Lipid Bilayer Using Free Energy Perturbation Method. Biophysical Chemistry. 2021; 277: 106662.

51. Lan J, Ge J, Yu J, et al. Structure of the SARS-CoV-2 spike receptor-binding domain bound to the ACE2 receptor. Nature. 2020; 581: 215–220. https://doi.org/10.1038/s41586-020-2180-5.

52. Humphrey W, Dalke A, Schulten K. VMD: visual molecular dynamics. J Mol Graph. 1996; 14: 33-38, 27-38.

53. Huang J, Rauscher S, Nawrocki G, Ran T, Feig M, de Groot BL, Grubmüller H, MacKerell, A.D. Jr. CHARMM36m: an improved force field for folded and intrinsically disordered proteins. Nat Methods. 2017; 14: 71-73.

54. Phillips JC, Braun R, Wang W, Gumbart J, Tajkhorshid E, Villa E, Chipot C, Skeel RD, Kalé L, Schulten K. Scalable molecular dynamics with NAMD. J Comput Chem. 2005; 26(16):1781-802. doi: 10.1002/jcc.20289. PMID: 16222654; PMCID: PMC2486339.

55. Huang J, Rauscher S, Nawrocki G, Ran T, Feig M, de Groot BL, Grubmüller H, MacKerell Jr. AD. CHARMM36m: an improved force field for folded and intrinsically disordered proteins. Nat Methods. 2017; 14(1): 71-73. doi: 10.1038/nmeth.4067. Epub 2016 Nov 7. PMID: 27819658; PMCID: PMC5199616.

56. Seeber M, Cecchini M, Rao F, Settanni G, Caflisch A. Wordom: a program for efficient analysis of molecular dynamics simulations. Bioinformatics. 2007; 23: 2625-2627.

57. Seeber M, Felline A, Raimondi F, Muff S, Friedman R, Rao F, ..., Fanelli F. Wordom: a user-friendly program for the analysis of molecular structures, trajectories, and free energy surfaces. J. Comput. Chem. 2011; 32:1183-1194.

58. Fisher N I, Lee A J. A Correlation Coefficient for Circular Data. Biometrika. 1983; 70: 327−332.

59. Fisher N. I. Statistical Analysis of Circular Data. Cambridge University Press: New York, 1993, p 277.

60. Mardia K V, Jupp P E. Directional Statistics. Wiley: Chichester, UK, 2000, p 429.

61. Zhang L, Buck M. Molecular Dynamics Simulations Reveal Isoform Specific Contact Dynamics Between the Plexin Rho GTPase Binding Domain (RBD) and Small Rho GTPases Rac1 and Rnd1. J. Phys. Chem. B. 2017; 121: 1485–1498.

62. Zhang L, Centa T, Buck M. Structure and dynamics analysis on plexin-B1 Rho-GTPase binding domain monomer and dimer. J. Phys. Chem. B. 2014; 118: 7302–11.

63. Sethi A, Eargle J, Black A, Luthey-Schulten Z. Dynamical Networks in tRNA:protein Complexes. Proc. Natl. Acad. Sci. U.S.A. 2009; 106: 6620−6625.

64. Glykos N. Software News and Updates. Carma: a Molecular Dynamics Analysis Program. J. Comput. Chem. 2006; 27: 1765−1768.

65. Frenkel D, Smit B. Understanding Molecular Simulation (Academic Press, San Diego). 2nd Ed. 2002.

66. Kirkwood J G. "Statistical mechanics of fluid mixtures". J Chem Phys. 1935; 3: 300–313.

67. Straatsma T P, McCammon J A. Multiconfiguration thermodynamic integration. J Chem Phys. 1991; 95: 1175–1188.

68. Zacharias M, Straatsma T P, McCammon J A. Separation-shifted scaling, a new scaling method for Lennard-Jones interactions in thermodynamic integration, J. Chem. Phys. 1994; 100, 9025.

69. Beutler T C, Mark A E, van Schaik R C, Gerber P R, van Gunsteren W F. Avoiding singularities and numerical instabilities in free energy calculations based on molecular simulations. Chem. Phys. Lett. 1994; 222: 529–539.

70. Pohorille A, Jarzynski C, Chipot C. Good practices in free-energy calculations. J Phys. Chem. B. 2010; 114: 10235–10253.

71. Andersen C A F, Palmer A G, Brunak S, Rost B. Continuum secondary structure captures protein flexibility. Structure. 2002; 10(2):175-84. doi: 10.1016/s0969-2126(02)00700-1.

72. Laskowski R A, Swindells M B. LigPlot+: multiple ligand-protein interaction diagrams for drug discovery. J. Chem. Inf. Model. 2011; 51: 2778-2786.

73. Laffeber C, Koning K de, Kanaar R, Lebbink JHG. Experimental Evidence for Enhanced Receptor Binding by Rapidly Spreading SARS-CoV-2 Variants. J Mol Biol. 2021; 433(15):167058. doi: 10.1016/j.jmb.2021.167058. Epub 2021 May 21. PMID: 34023401; PMCID: PMC8139174.

74. Barton MI, MacGowan SA, Kutuzov MA, Dushek O, Barton GJ, van der Merwe PA. Effects of common mutations in the SARS-CoV-2 Spike RBD and its ligand, the human ACE2 receptor on binding affinity and kinetics. Elife. 2021; 10:e70658. doi: 10.7554/eLife.70658. PMID: 34435953; PMCID: PMC8480977.

75. Tian F, Tong B, Sun L, Shi S, Zheng B, Wang Z, Dong X, Zheng P. N501Y mutation of spike protein in SARS-CoV-2 strengthens its binding to receptor ACE2. eLife. 2021; 10: e69091.

76. Wang WB, Liang Y, Jin YQ, Zhang J, Su JG, Li QM. E484K mutation in SARS-CoV-2 RBD enhances binding affinity with hACE2 but reduces interactions with neutralizing antibodies and nanobodies: Binding free energy calculation studies. J Mol Graph Model. 2021; 109:108035.

77. Wang Q, Ye S B, Zhou Z J, Li J Y, Lv J Z, Hu B, Yuan S, Qiu Y, Ge X Y. Key mutations on spike protein altering ACE2 receptor utilization and potentially expanding host range of emerging SARS-CoV-2 variants. J Med Virol. 2022; 95(1): e28116. https://doi.org/10.1002/jmv.28116.

78. Xu C, Wang Y, Liu C, Zhang C, Han W, Hong X, Wang Y, Hong Q, Wang S, Zhao Q, Wang Y, Yang Y, Chen K, Zheng W, Kong L, Wang F, Zuo Q, Huang Z, Cong Y. Conformational dynamics of SARS-CoV-2 trimeric spike glycoprotein in complex with receptor ACE2 revealed by cryo-EM. Science advances. 2021; 7(1): eabe5575. doi:10.1126/sciadv.abe5575.

Comparison of RBD DSSP

wildtype

Trimutant

RBD Mutants Bound with ACE2

a) wt, bound   (RES373)

b) wt, free   (RES372)

c) RBD-N501Y   (RES480)

d) RBD-trimutant   (RES501)