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Structure-based study of immune receptors as eligible binding targets of coronavirus SARS-CoV-2 spike protein



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ABSTRACT

One of the most important challenges in the battle against contagious SARS-CoV-2 is subtle identification of the virus pathogenesis. The broad range of COVID-19 clinical manifestations may indicate diversity of virus-host cells. Amongst key manifestations, especially in severe COVID-19 patients, reduction and/or exhaustion of lymphocytes, monocytes, basophils, and dendritic cells are seen.; therefore, it is required to recognize that how the virus infects the cells. Interestingly, angiotensin-converting enzyme 2 (ACE2) as the well-known receptor of SARS-CoV-2 is low or non-expressed in these cells. Using computational approach, several receptor candidates including leukocyte surface molecules and chemokine receptors that expressed in most lineages of immune cells were evaluated as the feasible receptor of spike receptor-binding domain (RBD) of SARS-CoV-2. The results revealed the higher binding affinity of CD26, CD2, CD56, CD7, CCR9, CD150, CD4, CD50, XCR1 and CD106 compared to ACE2. However, the modes of binding and amino acids involved in the interactions with the RBD domain of spike were various. Overall, the affinity of immune receptor candidates in binding to SARS-CoV-2 RBD may offer insight into the recognition of novel therapeutic targets in association with COVID-19.

1. Introduction

Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) belongs to β -coronavirus genus causes zoonotic diseases, including the latest pandemic that has been emerged in the world. Positive singlestranded RNA SARS-CoV-2 virus has four main structural proteins include helical nucleocapsid protein (N), envelope glycoprotein (E), membrane glycoprotein (M), and spike glycoprotein (S) [1]. The virus RNA polymerase induces numerous of recombination. Continuous mutations have made the virus susceptible to transmission from animal to human and human to human [2]. Several of these mutations are D614G, N501Y and, P681H in the spike of the virus that is responsible for binding to its receptor, which has expanded the ability to infect and transmit the virus between humans [3–5].

Enveloped viruses mostly use multiple entry pathways, which can be divided into non-specific and specific bindings. Non-specific bindings aggregate viruses on the cell's surface, while specific bindings trigger proteins involved in endocytosis [6]. The viruses through non-specific bindings are attracted on the host cells glycosaminoglycans (Chondroitin and heparan-sulfate proteoglycans). Following initial bindings, the viruses by specific receptors are led to signaling cascades that stimulate endocytosis and are required for virus internalization. These incidents eventually spread infection to other cells and initiate the host immune responses [7]. It has been shown that spike glycoprotein of SARS-CoV-2 virus interacts with heparan sulfate and vimentin that facilitate binding of virus on the cells surface [8,9].

SARS-CoV-2 enters human cells through binding to transmembrane protein angiotensin-converting enzyme 2 (ACE2) as the well-known receptor [10]. Also, the virus requires proteases such as transmembrane protease serine 2 (TMPRSS2) and furin to cleave spike protein into S1 and S2 subunits to membrane fusion and internalization [11]. ACE2 is highly expressed in nasopharynx, oral mucosa, intestines, kidney, testis, and moderate in lung, heart, liver, vasculature, and low or none in nervous and immune systems [12,13].

SARS-CoV-2 disadjusts immune responses and leads to excessive

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Abbr	eviati	ions
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Angiotensin-Converting Enzyme 2 (ACE-2)	
Receptor-Binding Domain (RBD)	
Severe Acute Respiratory Syndrome Coronavirus 2 (SARS-CoV-2)	
Transmembrane Protease Serine 2 (TMPRSS2)	
Monocyte Chemoattractant Protein-1 (MCP-1)	
Cytokine Storm (CS)	
Intensive Care Unit (ICU)	
Acute Respiratory Distress Syndrome (ARDS)	
Programmed Cell Death Protein 1 (PD-1)	
T cell Immunoglobulin Mucin-3 (TIM-3)	
Middle East respiratory syndrome (MERS)	
Dipeptidyl Peptidase 4 (DPP4)	
Protein Data Bank (PDB)	

Iterative Threading ASSEmbly Refinement (I-TASSER)						
Human Rhinoviruses (HRV)						
Consensus Prediction Of Interface Residues in Transient complexes						
(CPORT)						
Multiple Sequence Alignment (MSA)						
Markov Cluster (MCL)						
Cell Adhesion Molecules (CAMs)						
Leukocyte Functional Ag-2 (LFA-2)						
Super Immunoglobulin Family (SIgF)						
Signaling Lymphocytic Activation Molecule (SLAM)						
Intercellular Adhesion Molecule-1 (ICAM-1)						
vascular cell adhesion molecule-1 (VCAM-1)						
Neural Cell Adhesion Molecule (NCAM)						
Neuropilin-1 (NRP1)						
T Follicular Helper (TFH)						

Table 1

Proteins in the study.

	Uniprot ID	PDB ID	Method		Resolution		R-value free	Organism		
SARS-CoV-2- Spike	P0DTC2	6M0J.E.	X-RAY DIFFRACTION		2.45 Å		0.227	Homo sapiens		
ACE2	Q9BYF1	6M0J.A.	X-RAY DIFFRACTION		2.45 Å		0.227	Homo sapiens		
CCR5	P51681	6MEO.B	ELECTRON MICROSCOPY		3.90 Å		_	Homo sapiens		
CCR9	P51686	5LWE.A	X-RAY DIFFRACTION		2.80 Å		0.243	Homo sapiens		
CD4	P01730	5U1F.M	ELECTRON MICROSCOPY		6.80 Å		_	Homo sapiens		
CD54-ICAM1	P05362	1Z7Z.I	ELECTRON MICROSCOPY		8.00 Å		_	Homo sapiens		
CD26-DPP4	P27487	1WCY.A	X-RAY DIFFRACTI	ON	2.20 Å		0.265	Homo sapiens		
CD102-	P13598	1ZXQ.A	X-RAY DIFFRACTI	ON	2.20 Å		0.295	Homo sapiens		
ICAM2								*		
MERS-CoV-	K9N5Q8	6L8Q.B	X-RAY DIFFRACTION		3.10 Å		0.241	MERS coronavirus		
RBD										
VP1, VP2,	P03303	1K5M.ABC	X-RAY DIFFRACTI	ON	2.70 Å –		-	Rhinovirus B14		
VP3										
			Sequence ProSA (7-		I-Tasser (C- Ramachandran p			plot quality (%)		
	Uniprot	Template	Sequence	ProSA (Z-	I-Tasser (C-	Ramachano	lran plot quality (%)			
	Uniprot ID	Template PDB	Sequence identity (%)	ProSA (Z- score)	I-Tasser (C- score)	Ramachano Most	lran plot quality (%) Additionally	Generously	disallowed	
	Uniprot ID	Template PDB	Sequence identity (%)	ProSA (Z- score)	I-Tasser (C- score)	Ramachano Most favored	lran plot quality (%) Additionally allowed	Generously allowed	disallowed	
CD150-SLAM	Uniprot ID Q13291	Template PDB 1E07.A	Sequence identity (%) 28	ProSA (Z- score) -6.33	I-Tasser (C- score) -3.37	Ramachand Most favored 77.9	Iran plot quality (%) Additionally allowed 13.8	Generously allowed 4.8	disallowed	
CD150-SLAM CD2	Uniprot ID Q13291 P06729	Template PDB 1E07.A 1HNF.A	Sequence identity (%) 28 51	ProSA (Z- score) -6.33 -5.11	I-Tasser (C- score) -3.37 -2.08	Ramachand Most favored 77.9 88.2	Iran plot quality (%) Additionally allowed 13.8 10.4	Generously allowed 4.8 0.3	disallowed 3.4 1.0	
CD150-SLAM CD2 2B4-CD244	Uniprot ID Q13291 P06729 Q9BZW8	Template PDB 1E07.A 1HNF.A 1E07.A	Sequence identity (%) 28 51 25	ProSA (Z- score) -6.33 -5.11 -4.38	I-Tasser (C- score) -3.37 -2.08 -1.60	Ramachand Most favored 77.9 88.2 78.5	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5	Generously allowed 4.8 0.3 3.3	disallowed 3.4 1.0 2.7	
CD150-SLAM CD2 2B4-CD244 CD147-BSG	Uniprot ID Q13291 P06729 Q9BZW8 P35613	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A	Sequence identity (%) 28 51 25 47	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47	Ramachano Most favored 77.9 88.2 78.5 83.8	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9	Generously allowed 4.8 0.3 3.3 1.8	disallowed 3.4 1.0 2.7 1.5	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L	Sequence identity (%) 28 51 25 47 26	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98	Ramachano Most favored 77.9 88.2 78.5 83.8 79.2	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2	Generously allowed 4.8 0.3 3.3 1.8 5.1	disallowed 3.4 1.0 2.7 1.5 2.5	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4GZ9.A	Sequence identity (%) 28 51 25 47 26 91.5	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67	Ramachano Most favored 77.9 88.2 78.5 83.8 79.2 87.3	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4	disallowed 3.4 1.0 2.7 1.5 2.5 0.4	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4G29.A BIAA.A	Sequence identity (%) 28 51 25 47 26 91.5 24	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69	Ramachano Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0	tran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4 2.8	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM CD50-ICAM3	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591 P32942	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4GZ9.A BIAA.A 1Z7Z.I	Sequence identity (%) 28 51 25 47 26 91.5 24 52	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54 -7.56	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69 -0.68	Ramachand Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0 84.7	tran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7 11.5	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4 2.8 2.6	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5 1.3	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM CD50-ICAM3 CD242-	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591 P32942 Q14773	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4G29.A BIAA.A 1Z7Z.I 4OIA.A	Sequence identity (%) 28 51 25 47 26 91.5 24 52 37	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54 -7.56 -4.38	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69 -0.68 0.57	Ramachand Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0 84.7 89.8	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7 11.5 9.7	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4 2.8 2.6 0.0	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5 1.3 0.5	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM CD50-ICAM3 CD242- ICAM4	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591 P32942 Q14773	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4G29.A BIAA.A 1Z7Z.I 4OIA.A	Sequence identity (%) 28 51 25 47 26 91.5 24 52 37	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54 -7.56 -4.38	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69 -0.68 0.57	Ramachano Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0 84.7 89.8	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7 11.5 9.7	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4 2.8 2.6 0.0	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5 1.3 0.5	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM CD50-ICAM3 CD242- ICAM4 CD106-VCAM	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591 P32942 Q14773 P19320	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4GZ9.A BIAA.A 1Z7Z.I 4OIA.A 3DMK.A	Sequence identity (%) 28 51 25 47 26 91.5 24 52 37 23	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54 -7.56 -4.38 -6.08	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69 -0.68 0.57 0.05	Ramachand Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0 84.7 89.8 86.6	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7 11.5 9.7 10.5	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4 2.8 2.6 0.0 1.8	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5 1.3 0.5 1.1	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM CD50-ICAM3 CD242- ICAM4 CD106-VCAM CXCR6	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591 P32942 Q14773 P19320 O00574	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4GZ9.A BIAA.A 1Z7Z.I 4OIA.A 3DMK.A 6WWZ.R	Sequence identity (%) 28 51 25 47 26 91.5 24 52 37 23 34	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54 -7.56 -4.38 -6.08 -2.47	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69 -0.68 0.57 0.05 -0.87	Ramachand Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0 84.7 89.8 86.6 85.1	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7 11.5 9.7 10.5 10.5	Generously allowed 4.8 0.3 3.3 5.1 0.4 2.8 2.6 0.0 1.8 2.8	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5 1.3 0.5 1.1 1.6	
CD150-SLAM CD2 2B4-CD244 CD147-BSG CD7 CD304-NRP1 CD56-NCAM CD50-ICAM3 CD242- ICAM4 CD106-VCAM CXCR6 XCR1	Uniprot ID Q13291 P06729 Q9BZW8 P35613 P09564 O14786 P13591 P32942 Q14773 P19320 O00574 P46094	Template PDB 1E07.A 1HNF.A 1E07.A 3B5H.A 5YFI.L 4GZ9.A BIAA.A 1Z7Z.I 4OIA.A 3DMK.A 6WWZ.R 5WB1.A	Sequence identity (%) 28 51 25 47 26 91.5 24 52 37 23 34 28	ProSA (Z- score) -6.33 -5.11 -4.38 -3.39 -5.14 -9.06 -5.54 -7.56 -4.38 -6.08 -2.47 -2.26	I-Tasser (C- score) -3.37 -2.08 -1.60 -0.47 -2.98 -2.67 -0.69 -0.68 0.57 0.05 -0.87 -0.27	Ramachand Most favored 77.9 88.2 78.5 83.8 79.2 87.3 77.0 84.7 89.8 86.6 85.1 87.5	Iran plot quality (%) Additionally allowed 13.8 10.4 15.5 12.9 13.2 11.9 17.7 11.5 9.7 10.5 10.5 9.3	Generously allowed 4.8 0.3 3.3 1.8 5.1 0.4 2.8 2.6 0.0 1.8 2.6 0.0 1.8 2.8 1.6	disallowed 3.4 1.0 2.7 1.5 2.5 0.4 2.5 1.3 0.5 1.1 1.6 1.6	

inflammatory responses. Reduced CD4⁺ and CD8+T cells, NK, B cells (lymphopenia), monocytes, basophils and dendritic cells are vital experimental manifestations, especially in severe COVID-19 patients [14,15]. These reduced cell phenotypes were associated particularly with high IL-1 β , TNF- α , IL-2, IL-6, IL-8, IL-10, IL-18, IFN- γ , IP-10 (CXCL-10), and MCP-1 serum levels (cytokine storm) (CS)) in those that required ICU [16–18]. CS can induce acute respiratory distress syndrome (ARDS) that results in alveolar collapse due to loss of surfactant, less oxygen entering the bloodstream and more fluid entering the alveolar [19–21].

Further, among T cells, a decrease in the memory T cell population (CD45RO +) has been shown that weaken COVID-19 patients' immune memory and predispose them to other diseases [22]. Likewise, T cells in patients with COVID-19 exhibit exhaustion phenotypes and PD-1 and

TIM-3 levels are increased in the T cells subset. These functionally exhausted T cells may contribute to severe circumstances in COVID-19 patients [23].

Lymphopenia was also demonstrated in the Middle East respiratory syndrome (MERS) patients. MERS-CoV can directly infect human T lymphocytes and induce T-cell apoptosis via dipeptidyl peptidase 4 (DPP4 or CD 26), which is plenty found in lymphocytes [24]. There is a high homology between SARS-CoV-2, SARS-CoV, and MERS-CoV [25]. As lymphocytes do not express ACE2, it remains to be looked at whether novel receptors might mediate SARS-CoV-2 entry into T cells [26]. For this purpose, we computationally investigated several candidate receptors to predict possibility of SARS-CoV-2 virus entry to these cells. Briefly, the binding affinity and mode of interaction of immune receptors to the SARS-CoV-2 RBD were compared to ACE2 interactions





Fig. 1. A **ConSurf analysis of SARS-CoV-2 spike.** A) The sequence of RBD (319–541 in red box) that is colored by their conservation grades. (e: an exposed residue; b: a buried residue; f: a predicted functional residue; s: a predicte7d structural residue and yellow: insufficient data). B) The 3D structure of the RBD is presented using a space-cartoon model. The amino acids are colored by their conservation grades using the color-coding bar, with cyan-through-purple indicating variable-through-conserved. Positions, for which the inferred conservation level was assigned with low confidence, are marked with light yellow.

using homology modeling and molecular docking. One of the most advantages of this study is the identification of new candidate receptors that can be considered in novel therapeutic strategies.

2. Materials and methods

2.1. Sequences of proteins

To find the homologous structures of proteins, the retrieved amino acid sequences of proteins from Uniprot in canonical FASTA format were evaluated against Protein Data Bank (PDB) using NCBI Blast and HHpred (Table 1). HHpred performs the query-template alignment of proteins using PDB_mmCIF70 database [27].

2.2. Tertiary structure prediction and validation

Homology modeling was implemented as a comparative modeling method to build models of the 3D conformation of the proteins with no homologous structure. Using I-TASSER (Iterative Threading ASSEmbly Refinement), construction of full-length models started from multiple threading alignments to find structural templates from the PDB. Then, iterative fragment assembly simulations were followed by structure refinement to generate the 3D protein structures [28]. The PDB format of 2B4, CD2, CD7, CD50, CD106, CD147, CD150, CD242, CXCR6, XCR1 and NRP1 was obtained by uploading their amino acid sequence in FASTA format (Table 1). Eventually, the quality and reliability of the models were assessed by Protein Structure Analysis (ProSA) server and PROCHECK [29,30].

2.3. Identification of functional regions in SARS-CoV-2 spike

The biologically important regions of SARS-CoV-2 spike protein were evaluated using Consurf web server. To identify functional regions in protein, Consurf web server calculates the degree of conservation at each amino acid of protein based on the evolutionary relationships among the homologous sequences. Starting from either sequence or 3D structure, the server automatically collects homologues using multiple sequence alignment and reflects the evolutionary relations with color-code of cyan-through-purple in the range of 9 to 0 from highly conserved to the most variable [31,32]. The sequence of SARS-CoV-2 spike (Uniprot ID: PODTC2) and the 3D structure of RBD (PDB ID: 6MOJ.E.) were conducted to Consurf web server using the default parameters.



Fig. 2. Interaction mode of Ig-like domain of CD2, CD56 and CD150 with the RBD domain of SARS-CoV-2 spike. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) CD2/RBD-SARS-CoV-2, B) CD56/RBD-SARS-CoV-2 and C) CD150/RBD-SARS-CoV-2. RBD-SARS-CoV-2 is colored in cyan. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

2.4. Molecular docking

The molecular interaction of protein-protein was explored by an information-driven flexible docking approach using HADDOCK and a rigid body docking by ClusPro webserver [33,34]. The 3D structure of proteins (either available PDBs or the predicted models) was uploaded as the first molecule (receptor) and the 3D structure of RBD of SARS-CoV-2 as the second one (ligand). Further, the molecular interaction of the Middle East respiratory syndrome (MERS) spike to CD26 and capsid proteins (VP1, VP2, VP3) of human rhinoviruses (HRV) to

CD54 was evaluated to compare with the interactions of SARS-CoV-2 spike. The active residues of proteins were defined using CPORT (Consensus Prediction Of Interface Residues in Transient complexes) that predicts protein-protein interface residues [35]. Visualization and analysis of complexes of the top HADDOCK cluster was performed using PyMol and Ligplot + software, respectively [36,37]. The complexes with the lowest interaction energy and the highest interaction bonds (resulted from both docking webserver) were selected to calculate the binding affinity (Δ G) and dissociation constant (Kd) using PRODIGY server [38].



Fig. 3. Interaction mode of Ig-like domain of 2B4, CD102 and CD4 with the RBD domain of SARS-CoV-2 spike. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) 2B4/RBD-SARS-CoV-2, B) CD102/RBD-Sars-CoV-2 and C) CD4/RBD-SARS-CoV-2. RBD-SARS-CoV-2 is colored in cyan. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

2.5. Molecular dynamics simulation

The molecular dynamics simulations (MD) were implemented for 12 complexes in the study using GROMACS version 2020.3 with the AMBER99SB-ILDN force field on a system accelerated with NVIDIA GPU [39]. System preparation was performed by solvation of proteins with SPC/E (extended simple point charge) water model in a cubic box of 10 Å marginal radius. The system ionization and neutralization were made with Na⁺ and Cl⁻ ions. Energy minimization of the system was

performed using the steepest descent algorithm to remove any steric clashes. The system was equilibrated with a canonical ensemble of 500 ps of NVT ensemble (constant number of particles, volume, and temperature) at 300 K using the Berendsen thermostat, following an isothermal-isobaric NPT ensemble (constant number of particles, pressure, and temperature). Finally, MD simulation was run at 300 K for a period of 100 ns with 2 fs time step. The output trajectories were analyzed in terms of time-dependent behavior of the system by calculating the root mean square deviation (RMSD), root-mean-square



Fig. 4. Interaction mode of the extracellular domain of CD7, CD50 and CD106 with the RBD domain of SARS-CoV-2 spike. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) CD7/RBD-SARS-CoV-2, B) CD50/RBD-SARS-CoV-2 and C) CD106/RBD-SARS-CoV-2. RBD-SARS-CoV-2 is colored in cyan. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

fluctuation (RMSF), radius of gyration (RoG), solvent accessible surface area (SASA) and hydrogen bonds (H-bonds). The graphs were visualized using GraphPad Prism version 8.4.3 for Windows.

2.6. Analysis of protein network

The biological information of protein-protein associations were collected using STRING database (Search Tool for Retrieval of

Interacting Genes/Proteins) and Enrichr [40,41].

2.7. Multiple sequence alignment (MSA) and phylogenic study

To find conserved residues and regions in receptor proteins, Multiple Sequence Alignment (MSA) was carried out using Clustal Omega [42]. The reviewed sequences retrieved from Uniprot were subjected to MSA. The results of alignments were subjected to build the phylogenetic tree



Fig. 5. Interaction mode of the extracellular domain of CCR9, XCR1 and NRP1 with the RBD domain of SARS-CoV-2 spike. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) CCR9/RBD-SARS-CoV-2, B) XCR1/RBD-SARS-CoV-2 and C) NRP1/RBD-SARS-CoV-2. RBD-SARS-CoV-2 is colored in cyan. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

using MEGA7 [43].

3. Results

3.1. 3D structure of proteins

The sequence-template alignment using Blastp and HHpred resulted in finding 3D structure of SARS-CoV-2 RBD, ACE2, CCR5, CCR9, CD4, CD54, CD26, CD102, MERS-CoV-RBD, VP1, VP2 and VP3 of rhinovirus in PDB (6M0J.E., 6M0J.A., 6MEO.B, 5LWE.A., 5U1F.M., 1Z7Z.I., 1WCY. A, 1ZXQ.A, 6L8Q.B, 1K5M.A, 1K5M.B and 1K5M.C, respectively) (Table 1).

3.2. Homology modeling of the proteins

The five top models of 3D structure predicted by I-TASSER for each sequence were ranked by C-scores ranging from-5 to 2. The one with the highest C-score was selected as the best model for further analysis



Fig. 6. Interaction mode of the extracellular domain of CXCR6 and CCR5 with the RBD domain of SARS-CoV-2 spike. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) CXCR6/RBD-SARS-CoV-2, and B) CCR5/RBD-SARS-CoV-2. RBD-SARS-CoV-2 is colored in cyan. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

(data in Table 1). Firstly, there was a convenient consistency between the predicted structure and the secondary structure prediction. Secondly, the stereochemical features of the predicted structures were assessed using PROCHECK, and ProSA servers. Data are summarized in Table 1. ProSA Z-score for each protein indicated the overall model quality by evaluating the conformity of the models with the experimentally determined protein chains. Analysis of Ramachandran plot of the 3D structures from PROCHECK indicated that an acceptable percentage of the residues could be detected in the most favored areas of plot and allowed regions of it.

3.3. Evolutionary analysis of the sequence and 3D structure of SARS-CoV-2 spike

In ConSurf, the evolutionary rate of sequence and 3D structure of SARS-CoV-2 spike was estimated based on a specified probabilistic model of amino-acid substitutions. The conservation scores calculated by ConSurf were depicted in a coloring scheme indicating the relative measure of evolutionary conservation at each sequence site. The analysis revealed the functional (f), exposed (e), structural (s) and buried (b) residues of the protein. There are number of functional (highly conserved and exposed) and structural residues in the RBD of spike (319–541), where they are depicted by their conservation grades. The sequence analysis of SARS-CoV-2 spike (including RBD) and the colored

protein structure (generated by PyMol) are shown in Fig. 1.

3.4. Comparable interaction of receptors and SARS-CoV-2 spike in the complexes

The interaction mode of receptors and SARS-CoV-2 spike was studied using HADDOCK webserver. The following active interface amino acid residues: K417, Y421, G446, Y449, Y453, L455, F456, R457, K458, S459, E471, I472, Y473, Q474, A475, G476, S477, T478, P479, C480, N481, G482, V483, E484, G485, F486, N487, C488, Y489, F490, Q493, G496, Q498, T500, G502 and Y505 from the RBD domain of SARS-CoV-2 spike protein were defined to drive molecular docking. The topranking pose for docked complexes were selected from HADDOCK clusters with the lowest total intermolecular energies (Kcal/mol) and the least average pairwise backbone RMSD (Å) at their interface. The results including interacting residues of either Ig-like domain or extracellular domain of receptors and residues of RBD domain of SARS-CoV-2 spike are depicted in Figs. 2–9. Further, the number of hydrogen bonds, free energy of binding and the dissociation constant (Kd) value of complexes are summarized in Table 2.

3.5. The dynamics of proteins/RBD-spike interactions

Considering the results in Table 2, the 12 proteins/RBD-spike



Fig. 7. Interaction mode of the extracellular domain of CD54, CD242 and CD147 with the RBD domain of SARS-CoV-2 spike. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) CD54/RBD-SARS-CoV-2, B) CD242/RBD-SARS-CoV-2 and C) CD147/RBD-SARS-CoV-2. RBD-SARS-CoV-2 is colored in cyan. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

complexes at the top of the table were subjected to 100ns MD simulations using the GROMACS package. The MD trajectories were analyzed in terms of root mean square deviation (RMSD) of the structures during simulations. RMSD plots are depicted in Fig. 10. A. While the lowest fluctuations were related to CD4_J CD26, and XCR1 complexes, the RMSD of the complexes was in the range of 0.2–1 nm and showed the steady states at the final time of simulations.

Furthermore, the average fluctuation of each residue during the simulation was calculated as the root mean square fluctuation (RMSF) of

the CA atoms. The results were plotted for each complex to display residue fluctuations of RBD-spike (Fig. 10B). Analysis of RMSFs pattern revealed that the RBD-spike residues in the complexes had relatively similar fluctuations in the range of 0.05–0.3 nm (except for the C-terminal residues of RBD-spike with higher fluctuation up to 0.45 nm). The average fluctuation of each residue of CD26 in complex with either RBDspike of SARS-CoV-2 or MERS was evaluated during the simulation. The RMSF pattern of residues in the interface of CD26 (240–440) indicated different residue fluctuations (Fig. 11A), where CD26 residues in



Fig. 8. Interaction mode of the extracellular domain of CD54 with the domain of VP1, VP2 and VP3 of Rhinovirus. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) CD54/VP1, B) CD54/VP2 and C) CD54/VP3. CD54 is colored in blue. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and VPs, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

complex with MERS showed higher fluctuations. Fluctuations in the RMSF plot of the CD26/RBD-spike ranged from 0.1 to 0.4 nm and in the CD26/MERS plot ranged from 0.01 to 0.7 nm.

The radius of gyration determines the protein structure compactness, the higher values throughout the protein size range describe a less tight packing, whereas the lower radius of gyration indicates the tighter packing of proteins. While the compression fluctuations of the complexes during the simulation are limited, the lowest compactness was related to the CD56/RBD-spike complex with RoG of 2 nm and the other

complexes showed RoG in the range of 2.4-3.2 nm (Fig. 11B).

The density of hydrogen bonds between two different proteins at an interface plays a significant role in joining two different proteins together. Number of H-bonds at the interface of proteins and RBD-spike were plotted during the simulation that showed diverse range of less than 5 to more than 20 H-bonds. The distribution of amino acid composition at the protein–protein interfaces and the difference of amino acid side chains resulted in varying number of interactions. There was a relatively well agreement between number of H-bonds after



Fig. 9. Interaction mode of the extracellular domain of CD26 and ACE2 with the RBD domain of SARS-CoV-2 spike as well as CD26 with MERS. The surface representation of complexes that has resulted from molecular docking by HADDOCK. A) ACE2/RBD-SARS-CoV-2, B) CD26/RBD-SARS-CoV-2 and C) CD26/MERS. RBD-SARS-CoV-2 and MERS are colored in cyan and wheat. Close-up view of the stick representation of amino acids (AA) involved in the interactions of proteins that are colored in red (inside the box) and green (bold-italic/outside) for receptors and RBD-SARS-CoV-2/MERS, respectively. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

molecular docking and molecular dynamic simulation analysis (Table 3, Fig. 11C).

The solvent accessible surface area (SASA) was computed for the interface of proteins and RBD-spike to evaluate desolvation of the protein cavity and their binding affinity. Higher SASA values are defined as more exposed amino acids into solvent, whereas, lower scores indicate more buried residues in the protein. The SASA plot analysis revealed that complexes are classified in three range of SASA: ACE2/RBD-spike and CD56/RBD-spike showed the highest value (60–70 nm²), CD26/

MERS had SASA around 50 $\rm nm^2$ and the other complexes were in the range of 35–45 $\rm nm^2$ (Fig. 11D).

3.6. Protein-protein interaction network and KEGG enrichment analysis of human receptors

Interacting residues of the RBD domain of SARS-Cov-2 spike into residues of receptors were mapped to find the most common residues of the RBD domain (Fig. 12A). CD2 and NRP1 had the highest number of

Table 2

Molecular docking results of complexes.

Complexes	HADDOCK energy (Kcal/	PRODIGY		ClusPro		PRODIGY	
	mol)	∆G (Kcal∕ mol)	Kd (M) at 37.0 C	Cluster Members	Lowest Energy (Weighted Score)	∆G (Kcal∕ mol)	Kd (M) at 37.0 C
CD26/RBD-spike CD26/RBD- MERS	-339.4 -321.7	-16.3 -11.1	3.40E-11 1.60E-09	223 152	-3512.6 -1923.9	-16.5 -13.6	2.10E-12 2.40E-10
CD2/RBD-spike	-319.6	-16.2	3.70E-11	105	-2892.6	-17.3	6.80E-12
CD56/RBD-spike	-310	-15.1	2.10E-11	143	-2960.5	-14.3	8.00E-11
CD7/RBD-spike	-349.2	-15	2.60E-11	238	-3804.3	-14.2	1.00E-11
CCR9/RBD-spike	-329.9	-14.2	9.40E-11	253	-4031.5	-15.3	2.60E-12
CD150/RBD-	-337	-13.7	2.30E-10	219	-3266.8	-14.2	9.10E-11
spike							
CD4/RBD-spike	-295.2	-13.3	4.50E-10	185	-2092.4	-18.6	8.00E-10
CD50/RBD-spike	-272.6	-13.2	4.60E-10	177	-2407.2	-13.4	3.40E-10
XCR1/RBD-spike	-314	-13.1	5.80E-10	212	-3317.5	-18	1.90E-11
CD106/RBD-	-351.9	-12.9	7.70E-10	399	-4642.8	-14	1.30E-10
spike							
ACE2/RBD-spike	-393.6	-12.8	8.80E-10	358	-4587.4	-13.6	1.00E-11
CD54/RBD-spike	-248.5	-12.8	9.00E-10	186	-1952.3	-14.7	4.30E-10
CD54/VP1	-386.2	-12.6	1.40E-09	49	-1759.1	-16.7	1.60E-10
CD54/VP2	-240.1	-13.3	4.40E-10	93	-1882.1	-14.3	8.30E-10
CD54/VP3	-340.9	-13.7	2.00E-10	80	-2631.4	-16.2	3.60E-10
NRP1/RBD-spike	-382.3	-12.7	1.10E-09	216	-2682	-13.9	1.50E-10
CD242/RBD-	293.6	-12.6	1.40E-09	178	-2005.5	-12.7	1.00E-09
spike							
CCR5/RBD-spike	-222.8	-12.5	1.50E-09	87	-2433.2	-12	3.70E-09
CXCR6/RBD- spike	-330.1	-12.4	1.90E-09	93	-2599.2	-13.5	4.60E-10
CD147/RBD-	-235.9	-11.9	4.40E-09	145	-2488.1	-13.3	4.30E-10
cD102/RBD- spike	-275.5	-11.6	6.40E-09	298	-2766.9	-10.6	9.40E-10
2B4/RBD-spike	-202.5	-11.5	7.50E-09	146	-2496.6	-11.4	8.90E-09

interacting residues, though their interaction pattern was different. Interestingly, the variable and less protected amino acids of the RBD domain of spike were among the common amino acids involved in the interaction with the receptors rather than the more conserved amino acids. For example, K417, G446, Y449, S477, N481 and G485, which are less conserved, co-interact with the most receptors.

The full STRING network with a medium FDR stringency of 5 % and a medium confidence cutoff of 0.4 was created to find PPI network of above-mentioned receptors (Fig. 12B). STRING database predicted the functional interaction pattern of proteins. By applying the Markov Cluster algorithm (MCL) as clustering method (default: 3), proteins separated as DPP4 and ACE2 together, ICAM1-4, VCAM1, XCR1 and BSG in the next cluster and NRP1, NCAM1, CD244, SLAMF1, CD7, CD2, CD4, CCR5, CCR9 and CXCR6 in the third cluster. The experimentally interaction is predicted for CD4/CCR5 and CD4/CD2 (magenta line), whereas, the remaining partners showed text mining association in the form of light-yellow lines.

KEGG analysis revealed that the receptors were enriched in ten pathways including, Cytokine-cytokine receptor interaction, Chemokine signaling pathway, Hematopoietic cell lineage, Cell adhesion molecules (CAMs), Human immunodeficiency virus1 infection, Human T-cell leukemia virus1 infection, Renin-angiotensin system, Primary immunodeficiency, Intestinal immune network for IgA production and Antigen processing and presentation. The terms in which the receptors were enriched and ordered by p-value are listed in Fig. 12. C.

To trace evolutionary relation of proteins in terms of common domains, the receptor protein sequences were aligned and compared based on Neighbor Joining method with bootstrapping to provide confidence for tree topology (Fig. 12D). The percentage of replicate trees in which the associated taxa clustered together in the bootstrap values (1000 replicates) has been depicted next to the branches. Proteins in a common family (ICAM2/4 and ICAM1/3) showed higher score, no close and highscore evolutionary relationship was observed between receptor sequences.

4. Discussion

SARS-CoV-2 infection can result in a considerable reduction in circulating lymphocytes, NK, monocytes, basophils and dendritic cells, and proper immune potential in both severe and moderate patients. One reason for this reduction may be due to the direct entry of the virus into the cells. This infectivity occurs through the surface receptors of these cells such as ACE-2. Paradoxically, ACE-2 is not thoroughly expressed in the cells, the question remains to be answered whether other receptors are involved in virus entry [44]. To answer this question, we evaluated a number of candidate receptors that could facilitate virus entry into the cell.

Since the dominant technique that was implemented to evaluate the protein-protein interactions was molecular docking, there is a need to validate the docking programs. A common method for validation of the docking programs is re-docking the components of a known conformation and comparing to the co-crystallized structure. A program that results in returning the poses below an acceptable Root Mean Square Deviation (RMSD) value from the known conformation (usually 1.5 or 2 Å) is considered a successful docking program [45]. Likewise, RMSD of re-docking conformation of ACE2/spike from PDB: 6M0J and CD26/MERS from PDB: 6L8Q with both docking programs were calculated using align option in PyMol software. The RMSD of ACE2/spike complexes of HADDOCK and ClusPro aligned to 6M0J and CD26/MERS complexes aligned to 6L8Q were 0.921, 0.996, 0.908 and 0.942, respectively. Likewise, there was a well accordance in the results of complexes of both programs when evaluated in terms of the binding affinity and dissociation constant by PRODIGY.

In our study, CD26, which is the well-known receptor of the MERS-CoV virus and plays an important role in lymphocyte infection [24], has the strongest interaction with the SARS-CoV-2 virus spike. CD26 or dipeptidyl peptidase-4 (DPP4) is identified as a co-stimulator for T cell activation and expressed on T cell and also B cell, macrophages, NK, and thymocytes [46–48]. In consistent with the results of Li et al. that



Fig. 10. RMSD and RMSF plot of the complexes after 100 ns by MD simulations. A) RMSD of the backbone atoms of protein/RBD-spike complexes that showed finally the systems appeared to reach equilibrium after 100 ns. B) The RMSF plots of RBD-spike in complexes. Key residues of RBD-spike are depicted by arrow.

reported almost identical binding pattern of SARS-CoV-2-S RBD and MERS-CoV-S RBD with DPP4 using ZDOCK [49], HADDOCK outputs revealed the CD26 residues K267, R336, R317, and Q344 contributed significantly in the interaction with both RBDs. However, the binding affinity resulted from both molecular docking programs showed higher affinity of CD26/RBD-SARS-CoV-2 compared to MERS-CoV-S RBD/CD26.

It is proposed that cell adhesion molecules could be convenient candidates for virus binding considering weak physiological ligand binding affinities compared to the stronger virus binding affinity for the receptor [50]. CD2 or leukocyte functional Ag-2 (LFA-2) is an integral glycoprotein and a member of the super immunoglobulin family (SIgF) which is expressed on T, NK, NKT cells, and thymocytes. It acts as T cell adhesion molecule to target cell and antigen presenting cells, T cell activation, signal transduction, and regulation of cytolysis [51,52]. Other members of the CD2 family that could mention as potentially SARS-CoV-2 receptor include CD150 or signaling lymphocytic activation molecule (SLAM) and CD244 or 2B4 [53,54]. The binding affinity of CD2/spike was higher than ACE2/spike, though there was no current report of binding details of CD2 and spike. The SLAM receptor interacted with spike through immunoglobulin (Ig) domain and resulted in moderately higher binding affinity compared to ACE2/spike as



Fig. 11. RMSF, RoG, SASA, and H-bonds plot of the complexes after 100 ns by MD simulations. A) The RMSF plot of CD26 in complex with RBD-spike and MERS. B) Radius of gyration of complexes. C) The average number of H-bonds in the interface of proteins and RBD-spike. D) Solvent accessible surface area in the interface of proteins and RBD-spike.

cognitive SARS-CoV-2 RBD receptor. The result could be valuable knowing the capability of SLAM family receptors in binding to several morbilliviruses, such as the measles [55]. CD244/spike showed a different interaction pattern of SARS-CoV-2 RBD residues (G446, A475, G482, C488 and T500), though it has shared some common interacting amino acids (S477 and N481) from spike with the ACE2 receptor. Thus, the binding affinity of CD244/spike was slightly lower than ACE2/spike.

CD7 and CD4 from other SIgF members also showed promising results in the analysis for SARS-CoV-2 spike binding affinity. Despite of different binding mode, SARS-CoV-2 RBD could bind to CD7 and CD4 stronger (-15 and -13.3 kcal/mol, respectively) than ACE2/spike (-12.8 kcal/mol). CD4 as a co-receptor for MHC class II-restricted antigen-induced T cell activation and HIV receptor exists in T cells, innate lymphoid cells, monocytes, and macrophages [56]. CD7 as regulator of cytokine production and T cell activation is on the surface of T, NK cells and, thymocytes [57]. Recent studies reported reduction in T cells of COVID-19 cases [16,58,59]. Considering high binding affinity of CD4/spike could induce more curiosity to find the probable SARS-CoV-2 mechanism in avoiding T cell activation. However, negatively correlation of CD4⁺ T cells and level of cytokines (TNF- α , IL-6) has suggested as the potential cause of T cell depletion in COVID-19 patients [23].

Other candidates that can be imagine for the virus binding are cell adhesion molecules (CAM) such as ICAM-1 or CD54 at the surface of lymphocytes, endothelial cells and fibroblasts that also, is receptor of Rhinoviruses [60], ICAM-2 or CD102 on the endothelial cells, ICAM-3 or CD50 on lymphocytes, ICAM-4 or CD242 on the majority of cells specially, erythrocytes and VCAM-1 or CD106 on the surface of macrophages, dendritic cells, endothelium [61]. The affinity of ICAMs and ACE2 in binding to SARS-CoV-2 RBD was relatively in the same range. Interestingly, CD54 showed similar binding affinity to SARS-CoV-2 RBD as well as VP1, VP2 and VP3 of Rhinoviruses. The results reinforce the need to pay more attention to the role of these receptors in interacting with SARS-CoV-2.

Other adhesion molecules that can be thought as receptors for the virus include CD56 or neural cell adhesion molecule (NCAM) [62], Neuropilin-1 (NRP1), and Basigin (BSG or CD147). The first, being present in NK and brain cells, the second in T cells and neurons and the third in pericytes, endothelial cells and leukocytes. Actually, experimental studies proved the NRP1 and BSG as receptors for the SARS-CoV-2 virus [63–66] and Meplazumab (CD147-specific humanized monoclonal antibody) as a therapeutic agent [67]. Interacting residues of SARS-CoV-2 RBD were various in binding to NCAM, BSG and NRP1, though the affinity of BSG and NRP1 to RBD was nearly similar. Consistent with the results of Ulrich et al. binding affinity and residues involved in the coronavirus RBD interactions highlighted CD147 as a target for COVID-19 treatment [68].

Chemokine receptors, in addition to their primary role in cellular chemotaxis, are also known as receptors for viruses. Amongst, CXCR6

Table 3

Molecular docking and Molecular dynamic simulation results of complexes.

Complexes	HADDOCK RMSD (Å)	Number of H- bonds from docking	Average number of H- bonds from MD	SASA (nm/ S2/N)	RoG (nm)
CD26/RBD- spike	1.5	13	12	43	2.6
CD26/RBD- MERS	0.4	5	13	50	2.7
CD2/RBD-	0.8	17	9	44	2.85
CD56/RBD- spike	2.2	12	9	68	2.2
CD7/RBD- spike	1.8	11	10	38	2.4
CCR9/RBD- spike	0.8	13	10	43	3.0
CD150/ RBD- spike	0.4	9	9	39	2.9
CD4/RBD- spike	0.4	11	10	38	2.5
CD50/RBD- spike	2.3	6	5	45	2.75
XCR1/RBD- spike	1.9	16	15	37	3.1
CD106/ RBD- spike	1.2	15	14	36	2.5
ACE2/RBD- spike	1.5	13	14	65	2.8
CD54/RBD- spike	0.6	11	-	-	-
CD54/VP1	0.46	12	-	-	-
CD54/VP2	0.52	10	-	_	-
CD54/VP3	1.2	7	-	_	-
NRP1/RBD- spike	0.5	18	-	-	-
CD242/ RBD- spike	2.6	9	-	-	-
CCR5/RBD- spike	2.0	8	-	-	-
CXCR6/ RBD- spike	1.5	14	-	-	-
CD147/ RBD-	1.7	8	-	-	-
CD102/ RBD-	1.5	11	-	-	-
spike 2B4/RBD- spike	1.7	7	-	_	-

(CD186), CCXCR1 (XCR1), CCR9 (CDw199), and CCR5 (CD195) that are expressed in a broad span of immune cells have shown the eligible interactions with spike of virus in docking analysis. A recent study reported that treatment with Leronlimab (CCR5 blocking antibody) have displayed restoration of immune system and a significant decrease in SARS-CoV-2 plasma viremia [69]. In the current study, relatively different interface of SARS-CoV-2 RBD was resulted in binding to CXCR6, XCR1, CCR9, and CCR5, although similar binding affinity was obtained in the CXCR6, XCR1, and CCR5 complex analysis (-12.4, -13.1 and -12.5 kcal/mol, respectively). The binding affinity of CCR9 complex was higher (-14.2 kcal/mol) compared to ACE2/spike.

Considering the importance of the structural behavior of proteinprotein complexes, they were subjected to MD simulation to study the dynamic feature of a protein at atomic levels. MD provides insight into the conformational changes of proteins through time. RMSD, as the estimation of the average distance of a set of atoms in the input proteins for the same atoms during simulation, indicated the structural stability of complexes. Despite of various fluctuations, finally all complexes reached relatively steady state. The fluctuation pattern of the key residues of RBD-spike was approximately in the same range for all complexes, though CD26/MERS showed higher fluctuations compared to CD26/RBD-spike. The difference was consistent with the results of docking energy and binding affinity, where CD26/MERS complex showed smaller docking energy and Kd compared to CD26/RBD-spike and could be interpreted as the higher stability of CD26/RBD-spike. Further, the number of H-bonds in the interface of proteins and RBDspike was remarkably comparable to the LigPlot analysis of complexes and could be considered as the stability of complexes during the simulation.

Besides of ACE2 diverse expression in many cell types also, the results of our study indicate extent of spike SARS-CoV-2 virus interaction with the mentioned receptors, which may be related to the broad range of clinical manifestations such as vasculitis (VCAM-1 on the endothelium), and neuro-olfactory disorders (NRP-1, NCAM, CD-147 on the neurocytes) [70,71]. Likewise, infection of the immune cells with the SARS-CoV-2 virus can be important because they may be served as a "Trojan horse" to spread the virus throughout the human body.

The importance of the reduction of T Cells in COVID-19 is determined by the fact that despite of the antibody response and cytokines production by other immune cells in severe patients, they often have trouble in management of SARS-CoV-2, permitting the virus to proliferate in high levels. The innate immune cells respond by secreting still more cytokines and cause cytokine storm [72,73]. Furthermore, for effective antibody responses, bilateral cooperation of T and B lymphocytes in the germinal centers of the lymphatic tissue is required [74]. B cells within germinal centers shuffle Immunoglobulin (Ig) genes via somatic hyper mutation and generate Ig variants which afterward contend for T follicular helper (TFH) cell survival signal and as a result, B cells that generating low affinity and auto-reactive antibodies are omitted. This process promotes the affinity maturation of the antibodies generated by the B cells and facilitates differentiation of these B cells into plasma and memory B cells, essential cell types for durable antibody production, and impressive humoral immune responses [75,76]. Accordingly, the lack of functional T Cells has resulted in appearance of antibodies that are not neutralizing the virus and cannot prevent the progression of the disease or react against their self-cells. This mechanism is important because in the etiology of COVID-19 has shown that auto-antibodies, by attacking the blood vessels, cause clots, which ultimately leads to impaired blood flow and stroke. Overall, the affinity of several immune receptor candidates in binding to SARS-CoV-2 RBD was evaluated aiming to highlight potential roles in association with COVID-19. Further studies and experiments can improve the results of this research.

5. Conclusion

Considerable reduction of lymphocytes, monocytes, basophils, and dendritic cells in severe and moderate SARS-CoV-2 infections induced us to evaluate the immune receptors that may be potentially involved in virus entry. The higher binding affinity of several immune receptors to the SARS-CoV-2 RBD compared to ACE2 can justify the aim of the study to consider other receptors as therapeutic targets for meeting the challenge of COVID-19.

Author contributions were as following

Conception and design the study: Saeed Mobini, Milad Chizari, Elham Rismani.

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Fig. 12. Interaction map of the RBD domain of SARS-Cov-2 spike into receptors as well as PPI network analysis. A) Interacting residues on the basis of molecular docking results as well as conservation grade of residues of the RBD domain of spike. B) STRING PPI network of human proteins. C) Enrichment pathway analysis of PPI network in KEGG. D) Phylogenetic tree of protein sequences using Neighbor Joining method with bootstrapping to provide confidence for tree topology. The percentage of replicate trees in which the associated taxa clustered together in the bootstrap values (1000 replicates) has been depicted next to the branches.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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