

Bioinformatic prediction of immunodominant regions in spike protein for early diagnosis of the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2)

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Background. To contain the pandemics caused by SARS-CoV-2, early detection approaches with high accuracy and accessibility are critical. Generating an antigen-capture based detection system would be an ideal strategy complementing the current methods based on nucleic acids and antibody detection. The spike protein is found on the outside of virus particle and appropriate for antigen detection.

Methods. In this study, we utilized bioinformatics approaches to explore the immunodominant fragments on spike protein of SARS-CoV-2.

Results. The S1 subunit of spike protein was identified with higher sequence specificity. Additionally, glycosylation sites and high-frequency mutation sites on spike protein were circumvented in the antigen design. Three immunodominant fragments, Spike₅₆₋₉₄, Spike₁₉₉₋₂₆₄, and Spike₅₇₇₋₆₁₂, located at the S1 subunit were finally selected via bioinformatics analysis. All these fragments present qualified antigenicity, hydrophilicity, and surface accessibility. A recombinant antigen with a length of 194 amino acids (aa) consisting of the selected immunodominant fragments as well as a universal Th epitope was finally constructed.

Conclusion. The recombinant peptide encoded by the construct contains multiple immunodominant epitopes, which could stimulate strong immune response in mice and generate qualified antibodies for SARS-CoV-2 detection.

1 **Bioinformatic prediction of immunodominant regions in spike protein**
2 **for early diagnosis of the severe acute respiratory syndrome**
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4

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21

22 **Abstract**

23 **Background.** To contain the pandemics caused by SARS-CoV-2, early detection approaches
24 with high accuracy and accessibility are critical. Generating an antigen-capture based detection
25 system would be an ideal strategy complementing the current methods based on nucleic acids
26 and antibody detection. The spike protein is found on the outside of virus particle and
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35 recombinant antigen with a length of 194 amino acids (aa) consisting of the selected
36 immunodominant fragments as well as a universal Th epitope was finally constructed.

37 **Conclusion.** The recombinant peptide encoded by the construct contains multiple
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39 qualified antibodies for SARS-CoV-2 detection.

40 **Keywords:** SARS-CoV-2, Spike protein, Antigen-capture, Immunodominant fragments,
41 COVID-19

42 Introduction

43 The severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) is highly contagious
44 and has caused more than 44 million infection cases and over 1 million deaths
45 (<https://www.who.int/>), posing huge economic and social burden internationally [1,2]. The
46 reports of SARS-CoV-2 reinfection cases suggest that stronger international efforts are required
47 to prevent COVID-19 re-emergence in the future[3]. Nevertheless, we cannot exclude the
48 possibility that SARS-CoV-2 become a seasonal epidemic [4]. Even worse, the large amount of
49 asymptomatic infections greatly increases the difficulties of epidemic control [5]. To date, most
50 vaccines against SARS-CoV-2 are still in the stages of clinical trials and their efficacy is
51 uncertain (<https://www.who.int/>). Thus, early diagnosis of infected cases and population
52 screening would still be the focus now [6].

53 The real-time reverse transcriptase polymerase chain reaction(RT-PCR) and antibody-
54 capture serological tests are currently the main diagnostic methods for SARS-CoV-2 [7]. As the
55 golden standard, RT-PCR is highly reliable [8,9]. However, the implementation costs and
56 relatively cumbersome operation problems make it a big challenge for large population
57 screening[10]. The antibody-capture serological test is convenient, but seroconversion generally
58 occurs in the second or third week of illness. Therefore, it is not ideal for the early diagnosis of
59 infection [11-13]. The antigen-capture test is an alternative diagnostic method that relies on the
60 immunodetection of viral antigens in clinical samples. Accordingly, this method could be applied
61 for the detection of early infection no matter if the patient was asymptomatic or not [14].
62 Compared with RT-PCR based detection method, it is relatively inexpensive and can be used at

63 the point-of-care.

64 Rapid viral antigen detection has been successfully used for diagnosing respiratory viruses
65 such as influenza and respiratory syncytial viruses [15-23]. The sensitivity and specificity of the
66 antigen-capture detection system depend highly on the antigen employed to generate antibodies
67 [22]. The spike protein is one of the structural proteins of SARS-CoV-2, with the majority
68 located on the outside surface of the viral particles [24-26]. It has a 76.4% homology with the
69 spike protein of SARS-CoV. Sunwoo's study showed that the bi-specific spike protein derived
70 monoclonal antibody system exhibited excellent sensitivity in SARS-CoV detection [27]. The
71 virus infection is initiated by the interaction of spike protein receptor-binding domain (RBD) and
72 angiotensin converting enzyme 2 (ACE2) on host cells. It is widely accepted that the spike
73 protein is one of the earliest antigenic proteins recognized by the host immune system [28-32].
74 Nevertheless, the difficulties of using spike protein as antigen are also obvious. It is not easy to
75 express and purify the full-length spike protein [33]. In addition, the spike protein is highly
76 glycosylated [26] and prone to mutation[34], which would challenge the sensitivity of antigen-
77 capture based detection method. Hence, it is critical to truncating the glycosylation and mutation
78 sites on spike protein as much as possible in antigen design [33,35]. In fact, a study using the
79 truncated spike protein to detect SARS-CoV achieved a diagnostic sensitivity of >99% and a
80 specificity of 100% [36], which suggests that truncated spike protein of SARS-CoV-2 could also
81 be an appropriate candidate for the early diagnostic testing and screening of SARS-CoV-2. In
82 this study, we analyzed the spike protein via bioinformatics tools to obtain immunodominant
83 fragments. The predicted sequences were joined together as a novel antigen to immunize mice

84 for the preparation of specific antibodies. The whole flowchart of our work is depicted in Fig. 1.
85 Epitopes information presented by this work may aid in developing a promising antigen-capture
86 based detection system in pandemic surveillance and containment.

87 **Method**

88 **Data retrieval and sequence alignment**

89 Human coronavirus (HCoV) includes α -coronaviruses and β -coronaviruses. HCoV-229E
90 and HCoV-NL63 belong to the former, HCoV-HKU1, HCoV-OC43, the Middle East respiratory
91 syndrome-related coronavirus (MERS-CoV), SARS-CoV, and the SARS-CoV-2 belong to the
92 latter. We utilized the NCBI database to obtain sequences of all known human-related
93 coronaviruses spike protein in this study (HCoV-229E, HCoV-OC43, HCoV-NL63, HCoV-
94 HKU1, SARS-CoV, MERS-CoV, and SARS-CoV-2). Then, the Clustal Omega Server-Multiple
95 Sequence Alignment was used to analyze the downloaded sequences. Clustal Omega is a new
96 multiple sequence alignment program that uses HMM profile-profile techniques and seeds guide
97 trees to generate alignments between three or more sequences. In this study, we set our
98 parameters as default [37]. For further comparison between SARS-CoV-2 and SARS-CoV, we
99 exerted the EMBOSS Needle Server-Pairwise Sequence Alignment. Needleman-Wunsch
100 alignment algorithm supports this server to find the optimum alignment (including gaps) of two
101 sequences by reading two input sequences and translating their optimal global sequence
102 alignment to files [38]. It was also performed to compare major domains between SARS-CoV-2
103 and SARS-CoV.

104

105 **Linear B-cell epitope prediction**

106 Linear B-cell epitopes of the SARS-CoV-2 spike protein were calculated by ABCpred and
107 Bepipred v2.0 server. The ABCpred server predicts B-cell epitopes based on artificial neural
108 networks (machine-based technique). It grants a score for each predicted peptide. The peptides
109 with high scores are more likely to be effective epitopes. For ABCpred, we set a threshold of 0.8
110 to achieve a specificity of 95.50% and an accuracy of 65.37% for prediction. The window length
111 was set to 16 (the default window length) in this study [39]. The BepiPred v2.0 combines a
112 hidden Markov model and a propensity scale method. The score threshold for the epitope was set
113 to 0.5(the default value) to obtain a specificity of 57.16% and a sensitivity of 58.56% [40]. The
114 residue with scores above 0.5 was predicted to be part of an epitope.

115

116 **T-cell epitope prediction**

117 The free online service TepiTool server, provided by the Immune Epitope Database (IEDB),
118 was used to forecast epitopes binding to MHC-I and MHC-II molecules. TepiTool is a part of
119 IEDB, providing some top MHC binding prediction algorithms for many species such as
120 humans, mice, and so on [41]. For MHC-I binding epitopes, the mice were selected as host
121 species. And we selected alleles including H-2-Db, H-2-Dd, H-2-Kb, H-2-Kd, H-2-Kk, and H-2-
122 Ld for analysis. We took “IEDB recommended” as a prediction method and selected sections
123 with predicted consensus percentile rank ≤ 1 as predicted epitope. As the default prediction
124 method, “IEDB recommended” is based on the availability of predictors and previously observed
125 predictive performance. The purpose of this method was to ensure the best possible method for a

126 given MHC-I molecule [42]. For MHC-II binding epitopes, we selected the mice as host species
127 and adopted alleles including H2-IAb, H2-IAc, and H2-IEd for analysis. As same to MHC-I, we
128 chose “IEDB recommended” as a prediction method, and peptides with predicted consensus
129 percentile rank ≤ 10 were determined as epitopes. As above, this approach chose the best suitable
130 method for a given MHC-II molecule according to evaluations by the IEDB team and
131 bioinformatics community [43,44].

132

133 **Profiling and evaluation of selected fragments**

134 The secondary structure of the SARS-CoV-2 spike protein (PDB ID: 6VSB chain B) was
135 calculated by the PyMOL molecular graphics system using the SSP algorithm. PyMOL
136 (<http://www.pymol.org>) is a python-based tool, which is widely used when visualization of
137 macromolecules is needed. Most tasks of representations could be achieved by PyMOL [45].
138 Vaxijen2.0 server was utilized to analyze the antigenicity of epitopes and selected fragments.
139 VaxiJen was the first server for alignment-independent prediction of protective antigens, which
140 is based on auto cross-covariance (ACC) transformation of protein sequences into uniform
141 vectors of principal amino acid properties. A default threshold of 0.4 was set and the prediction
142 accuracy of this server is 70% to 89 % [46]. The hydrophilicity of the selected fragment was
143 analyzed by the online server ProtScale. ProtScale was one of the protein identification and
144 analysis tools in the ExpASy server, which could calculate hydrophilicity based on amino acid
145 scales [47]. Surface accessibility of predicted fragments was evaluated with the NetsurfP, an
146 online server predicting the surface accessibility and secondary structure of amino sequence [48].

147 Critical features such as allergenicity and toxicity were evaluated by online server AllerTOP v2.0
148 [49] and ToxinPred [50]. PyMOL was used to visualize the selected fragments on the SARS-
149 CoV-2 spike protein trimer. At last, we run a Protein BLAST search providing by NCBI
150 database for the eventual construct, to check the possibility of cross-reactivity with other human
151 and mouse proteins.

152

153 **Results**

154 **Sequence alignment of spike protein in different coronaviruses**

155 Coronaviruses had four genera composed of alpha-, beta-, gamma- and delta-coronaviruses.
156 Among them, alpha- and beta- genera could infect humans. Including SARS-CoV-2, seven
157 coronaviruses are known to infect humans (HCoV-229E, HCoV-OC43, HCoV-NL63, HCoV-
158 HKU1, SARS-CoV, MERS-CoV, and SARS-CoV-2) [51,52]. Sequences of these viruses were
159 obtained from the NCBI database, of which accession numbers were presented in Fig. 2A. We
160 performed sequence alignment to determine the evolutionary relationships between SARS-CoV-
161 2 and previously identified coronaviruses. According to the results of sequence alignment and
162 phylogenetic tree (Fig. 2A, Fig. 2B), SARS-CoV is the closest virus to SARS-CoV-2 among the
163 seven HCoVs, exhibiting a 77.46% sequence identity. To better understand the divergence of
164 spike protein sequences between SARS-CoV-2 and SARS-CoV, we further analyzed the
165 sequences of several major domains. Results showed that the S2 subunit was the most conserved
166 domain with a 90.0% identity. While RBM and NTD domains, both located in the S1 subunit,
167 only exhibited 49.3% and 50.0% identity respectively (Fig. 2C, Fig. 2D). Hence, we chose the

168 S1 subunit (amino acid 1-685) for the subsequent bioinformatics analysis as its potentially higher
169 specificity.

170

171 **Linear B-cell epitope prediction of S1 subunit in SARS-CoV-2 spike** 172 **protein**

173 The B-cell epitope is a surface accessible cluster of amino acids, which could be
174 recognized by secreted antibodies or B-cell receptors. It could elicit humoral immune response
175 [53]. The immunodominant fragments should contain high-quality linear B-cell epitopes to
176 stimulate antibody production effectively. The sequence of SARS-CoV-2 S1 subunit was
177 evaluated through ABCpred and BepiPred-2.0. The antigenicity was calculated via VaxiJen v2.0
178 with the given cutoff of ≥ 0.40 . A total of 31 peptides were identified by the ABCpred algorithm
179 (threshold set at 0.8, Table S1). For the Bepipred v2.0 server, we set a threshold of 0.5, and 14
180 epitopes were forecasted (Table S2). After antigenicity evaluation, 19 and 9 potential linear B-
181 cell epitopes predicted by the ABCpred server and BepiPred v2.0 server were obtained
182 respectively (Table 1). The area predicted by both bioinformatics programs is more likely to be
183 an epitope recognized *in vivo*. After mapping the positions of peptides identified by these
184 servers, 3 regions containing predicted epitopes were obtained. These regions could be
185 preliminarily considered as candidates for immunodominant fragments (Fig. 3, Table 2).

186

187 **Murine T-cell epitope prediction of S1 subunit in SARS-CoV-2 spike** 188 **protein**

189 Helper T cells (Th) recognize antigen peptides presented by MHC-II molecules and
190 facilitate the humoral immune response [54,55]. Therefore, the immunodominant fragments had
191 better contain MHC-II binding epitopes as well as linear B-cell epitopes. The S1 subunit was
192 selected for the prediction of T-cell epitopes. We utilized the TepiTool server to forecast MHC-I
193 and MHC-II binding epitopes. A total of 35 MHC-I binding epitopes was predicted (Table S3),
194 and 27 peptides were identified as MHC-II binding epitopes (Table S4). The antigenicity of these
195 peptides was calculated via Vaxijen 2.0 server (Table 3). Combined with the MHC-II epitopes
196 prediction results, the candidate immunodominant fragments were adjusted (Fig.4). Compared
197 with the preliminary candidate immunodominant fragments screened according to the linear B-
198 cell epitope prediction, we added the Spike₁₄₋₃₄ fragment into consideration because it contains a
199 linear B epitope and an MHC-II binding epitope, both of which had high antigenicity scores
200 (Table 4).

201

202 **Immunodominant fragments refinement according to the** 203 **glycosylation site distribution, mutation site distribution, and** 204 **secondary structure**

205 A profile of 24 glycosylation sites of SARS-CoV-2 spike protein has been reported [56].
206 Since glycans could hinder the recognition of antigens by shielding the residues [57], which
207 would affect the performance of antigen detection. Thus, when selected the immunodominant
208 fragments, we circumvent glycosylation sites as much as possible. According to the study of Asif
209 Shajahan et al, 15 glycosylation sites were located in the S1 subunit of the spike. Among them,

210 sites Asn¹⁷ and Asn⁶⁰³ were completely unoccupied[56]. Hence, the fragments in this study were
211 adjusted to Spike₁₄₋₃₄, Spike₄₉₋₁₀₁, Spike₁₉₉₋₂₆₁, and Spike₅₈₃₋₆₂₀. To retain some epitopes with high
212 antigenicity, the fragments inevitably contained 3 glycosylation sites.

213 Rapid transmission of COVID-19 provides the SARS-CoV-2 with substantial opportunities
214 for natural selection and mutations. To ensure the stability of the detection effect, the
215 immunodominant fragments were modified to avoid high-frequency mutation sites[58]. Spike₁₄₋₃₄
216 were excluded for containing four high-frequency mutation sites. Fragment Spike₄₉₋₁₀₁ was
217 adjusted to Spike₅₆₋₉₂, and fragment Spike₅₈₃₋₆₂₀ was adjusted to Spike₅₈₃₋₆₀₉. By adjusting the
218 fragments, we avoided in total of 8 high-frequency mutation sites. The adjusted fragments
219 contain none of the above high-frequency mutation sites, which could be beneficial to future
220 detection.

221 We use PyMOL to present the secondary structure of the spike protein (PDB ID: 6VSB
222) (Fig. S1). Attention was paid to maintaining the integrity of the secondary structure of the start
223 and end positions of fragments. Hence, the immunodominant fragments were finally adjusted to
224 Spike₅₆₋₉₄, Spike₁₉₉₋₂₆₄, and Spike₅₇₇₋₆₁₂. The epitopes and glycosylation sites contained in the
225 selected immunodominant fragments were displayed in Fig. 5.

226

227 **Profiling, evaluation, and visualization of selected immunodominant** 228 **fragments**

229 To further evaluate the antibody binding potentiality of these antigenic regions, several key
230 features of the selected fragments such as antigenicity, hydrophilicity, surface accessibility,

231 toxicity, and allergenicity were analyzed (Table 5). We used the full-length of subunit 1 spike
232 protein for computation of hydrophilicity and surface accessibility and analyzed antigenicity,
233 toxicity, and allergenicity using the sequence of selected fragments. Three fragments presented
234 relatively moderate hydrophilicity and surface accessibility. The proportion of hydrophilic amino
235 acids to the total length of the fragments Spike₅₆₋₉₄, Spike₁₉₉₋₂₆₄, and Spike₅₇₇₋₆₁₂ are 19/39, 30/66,
236 12/36 respectively. The surface accessibility of these fragments calculated by the online server
237 was shown in Table 5.

238 The antigenicity of the selected fragments was tested by the Vaxijen2.0 server. The toxicity
239 of the selected fragments was examined by ToxinPred, and no fragment was predicted to be
240 toxic. We assessed the allergenicity via the AllerTOP v2.0 server. Only fragment Spike₅₇₇₋₆₁₂ was
241 predicted to be a probable allergen. Attention should be paid to monitor potential allergic
242 reactions when injecting the recombinant protein into mice in further experiments. And the
243 selected fragments were presented as the sphere in the trimer structure (Fig. 6). Next, we
244 searched the sequence of the selected fragments and the epitopes in the IEDB server to determine
245 whether they were experimentally tested. None of the fragments contain experimentally tested
246 epitopes on the IEDB server, which also indicated that the fragments are highly specific for
247 SARS-CoV-2.

248

249 **Immunodominant fragments based recombinant antigen design**

250 Three immunodominant fragments embody several linear B-cell epitopes, MHC-I binding,
251 and MHC-II binding T-cell epitopes were selected. As a universal Th epitope, the PAN DR

252 epitope [PADRE(AKFVAAWTLKAAA)] can activate CD4+ cells, enhance helper T cell
253 activity, and assist activation of humoral immunity [59]. It was added into the construction
254 aiming to boost helper T cell activity [60,61]. (GGGGS)_n is a widely used flexible linker with the
255 function of segmenting protein fragments, maintaining biological activity, maintaining protein
256 conformation, and promoting protein expression [62]. Finally, we combined the fragments and a
257 PADRE epitope by linker peptide (GGGGS)₂ and (GGGGS)₃ [62](Fig. 7). The predicted
258 antigenicity of the final construct (194 aa) was 0.5690 (Table 6). A protein BLAST for the final
259 construct was conducted to evaluate the possibility of cross-reactivity. The blast results
260 suggested little similarity between the construct and any human or mice proteins (data not
261 shown).

262

263

264 Discussion

265 In this study, we explored the immunodominant fragments within the S1 subunit of the
266 SARS-CoV-2 spike protein. The final construct consists of three immunodominant fragments
267 Spike₅₆₋₉₄, Spike₁₉₉₋₂₆₄, Spike₅₇₇₋₆₁₂, and a PADRE epitope. After expression and purification, the
268 recombinant antigen will be used to immunize mice and qualified antibody generated could be
269 applied for developing an antigen-capture based detection system.

270 The antibody-based antigen capturing method is user-friendly, time-saving, and economical.
271 Thus, it is an ideal complementary detection strategy especially for early diagnosis and large
272 population screening. The monoclonal antibodies against SARS-CoV have been successfully

273 applied in the immunological antigen-detection of SARS-CoV [14]. Accordingly, we explored
274 the immunodominant fragments on the spike protein of SARS-CoV-2, which would provide aid
275 in developing an accurate and fast antigen-capture based early detection system for SARS-CoV-
276 2.

277 We selected the S1 subunit for immunodominant fragments screening after divergence
278 analysis. It had been reported that an S1 antigen-detected assay of SARS-CoV could detect the
279 virus as soon as the infection occurs [27]. Jong-Hwan Lee *et al.* designed a method which
280 captures and detects spike protein S1 subunit of SARS-CoV-2 using ACE2 receptor and S1-
281 mAb[63], which suggests that it is appropriate to use the S1 subunit for a specific and early
282 diagnosis of SARS-CoV-2. Three immunodominant fragments (Spike₅₆₋₉₄, Spike₁₉₉₋₂₆₄, and
283 Spike₅₇₇₋₆₁₂) were obtained in our study. These sequences will be joined to construct recombinant
284 peptides in the following step. The results of sequence BLAST suggested that these selected
285 fragments have a high specificity. Instead of using inactivated full-length spike protein, we
286 designed a novel recombinant protein construct which increased sequence specificity as well as
287 circumvented mutation sites and glycosylation sites. As the antigen design is totally based on
288 bioinformatics study, the exact ability of the selected fragments to produce qualified antibodies
289 for virus detection has yet to be identified by experiments.

290 Noticeably, the spike protein of SARS-CoV-2 is heavily glycosylated. Glycans could shield
291 epitopes during antibody recognition, and it may hinder the detection of viral proteins [56]. The
292 spike protein possessed 22 potential N-glycosylation sites along with two O-glycosylation sites.
293 Among 22 potential sites, about 17 N-glycosylation sites were found occupied [56]. We

294 deliberately circumnavigated glycosylation sites when selecting immunodominant fragments, but
295 several sites are preserved because they are located on the predicted epitopes. The three selected
296 fragments in this study still contain 4 glycosylation sites: Fragment Spike₅₇₇₋₆₁₂ does not contain
297 glycosylation sites. N-glycosylation sites Asn⁶¹ and Asn⁷⁴ located at Spike₅₆₋₉₄, and Asn²³⁴
298 located at Spike₁₉₉₋₂₆₄. To retain some epitopes with high antigenicity, the fragments inevitably
299 contained 3 glycosylation sites. In case these glycosylation sites do affect the diagnostic
300 performance, an additional deglycosylation step by N-glycanase could be applied for the test
301 specimens [64], which is a simple and efficient method for deglycosylation and is widely used in
302 multiple types of research [65-68]. Alternatively, a eukaryotic expressing system could be
303 employed to mimic the antigen presented in human cells.

304 Coronaviruses have the ability to correct errors during replication and recombination [58].
305 while the SARS-CoV-2 genome still presents many mutations. Many of these mutations could
306 spoil the proceeding effort in the development of diagnostic tests [69]. For diagnostic tests, the
307 mutations on the antigen are roadblocks in developing effective diagnostic tests against COVID-
308 19. In this study, we circumvented high-frequency mutation sites when selecting antigen
309 fragments. In addition, our fragments also avoided RBD regions which are prone to mutation
310 [69]. The construct finally built contained no high-frequency mutation.

311 To date, several studies using predictive algorithms to analyze SARS-CoV-2 have been
312 reported. However, most of them intended to design vaccines and focused on the homology part
313 of the virus sequence[70-72]. On the contrary, we aimed to detect the virus, more attention was
314 paid to the sections with large variability. Therefore, our immunodominant fragments are more

315 specific. In addition, unlike vaccine researches using humans as host species, we predicted
316 murine MHC-II binding T-cell epitopes to make sure the fragments could trigger a humoral
317 immune response in mice. Nevertheless, our conclusion is based on *in silico* calculations. The
318 efficiency needs to be evaluated by *in vitro* and *in vivo* experiments.

319 **Conclusion**

320 Through bioinformatics analysis, three immunodominant fragments were obtained in the
321 present study. After connected by flexible linkers, we acquired a final recombinant peptide with
322 a 194 aa length. It was predicted to have high antigenicity and possess specificity for SARS-
323 CoV-2. Our next move is to express and purify the recombinant protein in a suitable expression
324 system, followed by immunizing the mice with purified immunogen to obtain specific
325 antibodies. The present study would provide aid in the development of an antigen-capture based
326 detection system.

327

328 **Ethics approval and consent to participate**

329 Not applicable

330

331 **Author Contributions:**

332 **Siqi Zhuang:** Conception; Methodology; Software operation; Data analysis; Visualization;
333 Writing of the manuscript; Review and modification of the manuscript.

334 **Hongzhi Chen:** Conception; Methodology; Visualization; Review and modification of the
335 manuscript; Supervision and instruction; Project administration.

336 **Lingli Tang:** Conception; Methodology; Visualization; Review and modification of the
337 manuscript; Supervision and instruction; Project administration; Funding acquisition.

338 **Yufeng Dai:** Visualization; Review and modification of the manuscript.

339 All authors have read and agreed to the published version of the manuscript.

340

341

342 **Appendix A. Supplementary data**

343 The following are the supplementary data to this article:

344 **Fig. S1.** The secondary structure presentation of SARS-CoV2 spike using PyMOL server

345 **Table S1.** Linear B-cell epitopes predicted by the ABCpred server along with their position,
346 sequence, ABCpred prediction score, and antigenicity score.

347 **Table S2.** Linear B-cell epitopes predicted by the Bepipred v2.0 server along with their position,
348 sequence, and antigenicity score.

349 **Table S3.** MHC-I binding epitopes predicted by TepiTool along with their position, sequence,
350 allele, and antigenicity score.

351 **Table S4.** MHC-II binding epitopes predicted by TepiTool along with their position, sequence,
352 allele, and antigenicity score.

353

354 **Reference**

355

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545

Figure 1

Work flow chart

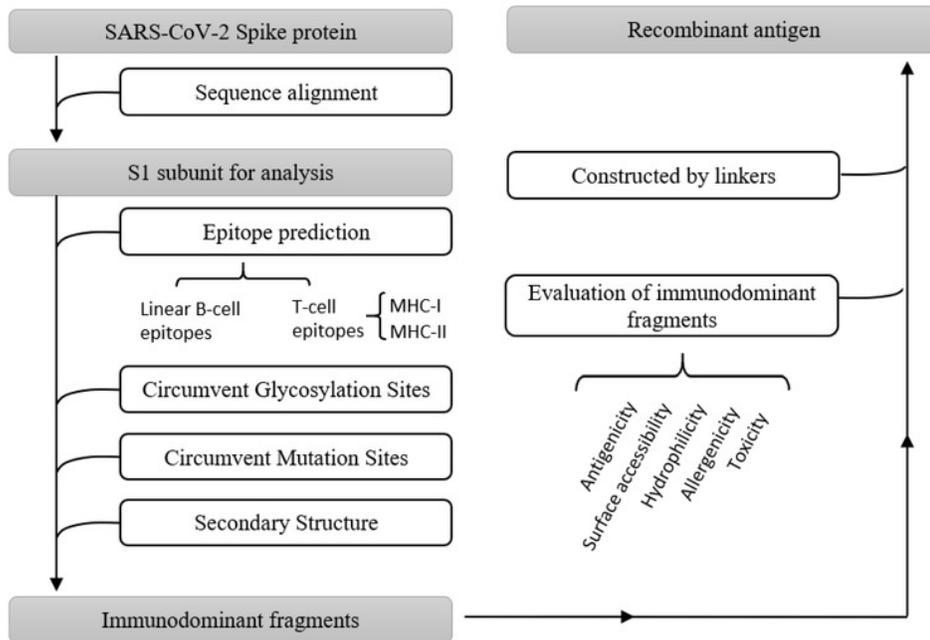


Fig. 1. Work flow chart

Figure 2

Sequence alignment results of spike protein

A. Accession IDs and sequence identities of selected coronavirus spike protein. B. Phylogenetic tree of spike proteins among selected coronavirus. C. Sequence identity of major domains in spike protein between SARS-CoV-2 and SARS-CoV. D. Sequence identity of domains in SARS-CoV-2 and SARS-CoV reflected by colors. From red to green, the color changing represents the sequence identity from high to low.

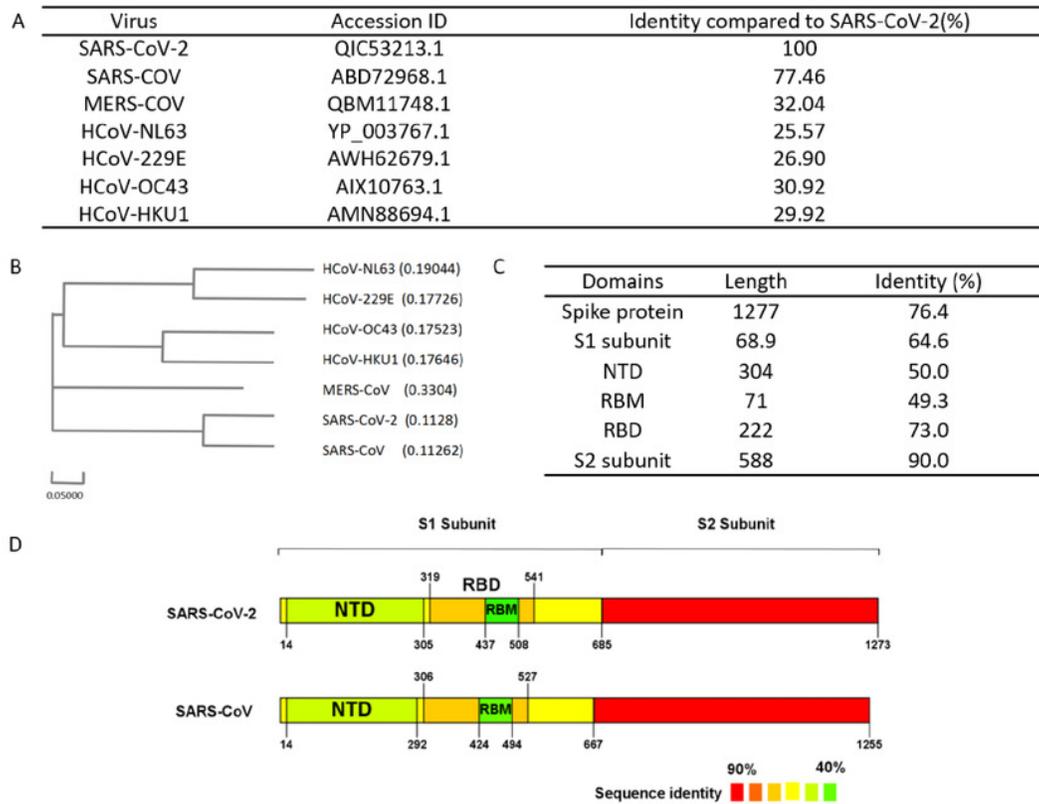


Fig. 2. Sequence alignment results of spike protein

- A.** Accession IDs and sequence identities of selected coronavirus spike protein.
- B.** Phylogenetic tree of spike proteins among selected coronavirus.
- C.** Sequence identity of major domains in spike protein between SARS-CoV-2 and SARS-CoV.
- D.** Sequence identity of domains in SARS-CoV-2 and SARS-CoV reflected by colors. From red to green, the color changing represents the sequence identity from high to low.

Figure 3

Preliminary immunodominant fragments based on B-cell epitope prediction results

The black squares represent epitopes predicted by ABCpred server, the black frames represent epitopes predicted by Bepipred v2.0 server, and the black lines with numbers on both ends represent the preliminary candidate immunodominant fragments.

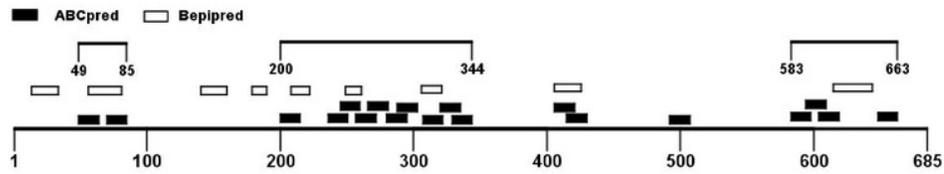


Fig. 3. Preliminary immunodominant fragments based on B-cell epitope prediction results

The black squares represent epitopes predicted by ABCpred server, the black frames represent epitopes predicted by Bepipred v2.0 server, and the black lines with numbers on both ends represent the preliminary candidate immunodominant fragments.

Figure 4

Adjusted candidate immunodominant fragments according to MHC-II T-cell epitope prediction results

The black squares represent epitopes predicted by ABCpred server, and the black frames represent epitopes predicted by Bepipred v2.0 server. The red frames denote MHC-II binding epitopes. The black lines with numbers on both ends represent the adjusted candidate fragments.

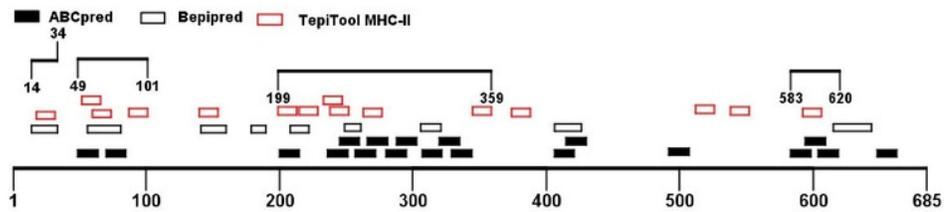


Fig. 4. Adjusted candidate immunodominant fragments according to MHC-II T-cell epitope prediction results

The black squares represent epitopes predicted by ABCpred server, and the black frames represent epitopes predicted by Bepipred v2.0 server. The red frames denote MHC-II binding epitopes. The black lines with numbers on both ends represent the adjusted candidate fragments.

Figure 5

The epitopes and glycosylation sites on the selected immunodominant fragments

The black squares represent epitopes predicted by ABCpred server, and the black frames represent epitopes predicted by Bepipred v2.0 server. The red squares represent MHC-I binding epitopes, and the red frames represent MHC-II binding epitopes. The grey squares means occupied glycosylation sites contained in the selected fragments.

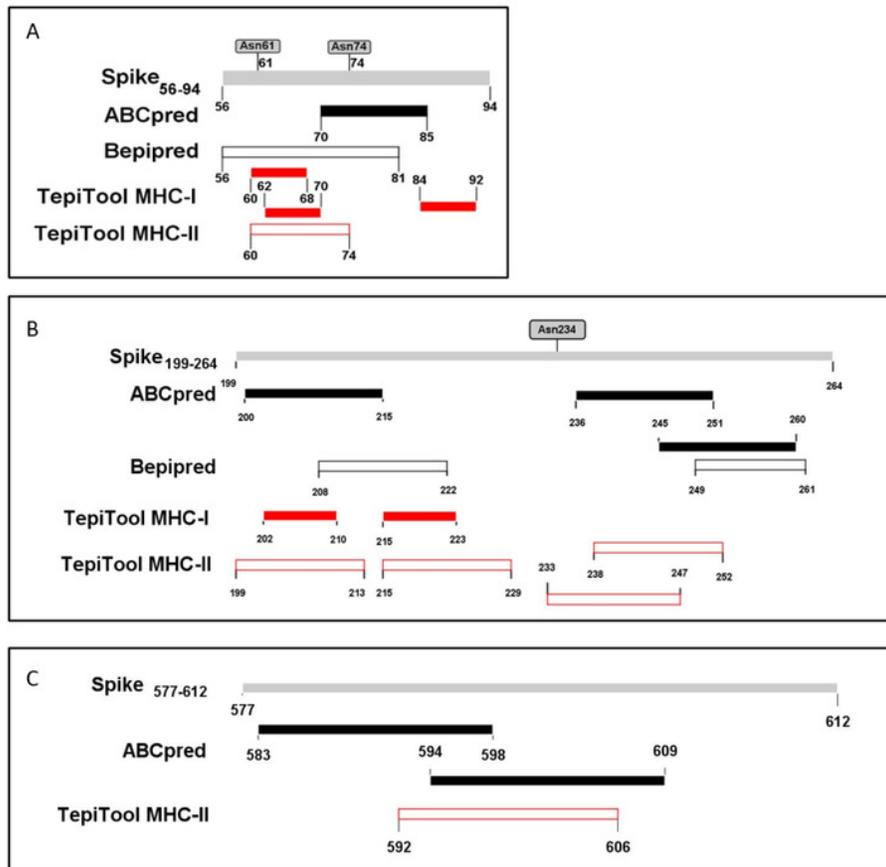


Fig. 5. The epitopes and glycosylation sites on the selected immunodominant fragments

The black squares represent epitopes predicted by ABCpred server, and the black frames represent epitopes predicted by Bepipred v2.0 server. The red squares represent MHC-I binding epitopes, and the red frames represent MHC-II binding epitopes. The grey squares means occupied glycosylation sites contained in the selected fragments.

Figure 6

Selected immunodominant fragments presented as spheres in the trimer structure of spike protein viewed by PyMOL

Selected fragments were presented as red spheres, green cartoons denote unselected sections. A, B, and C denote fragments Spike56-94, Spike199-264, and Spike577-612 respectively.

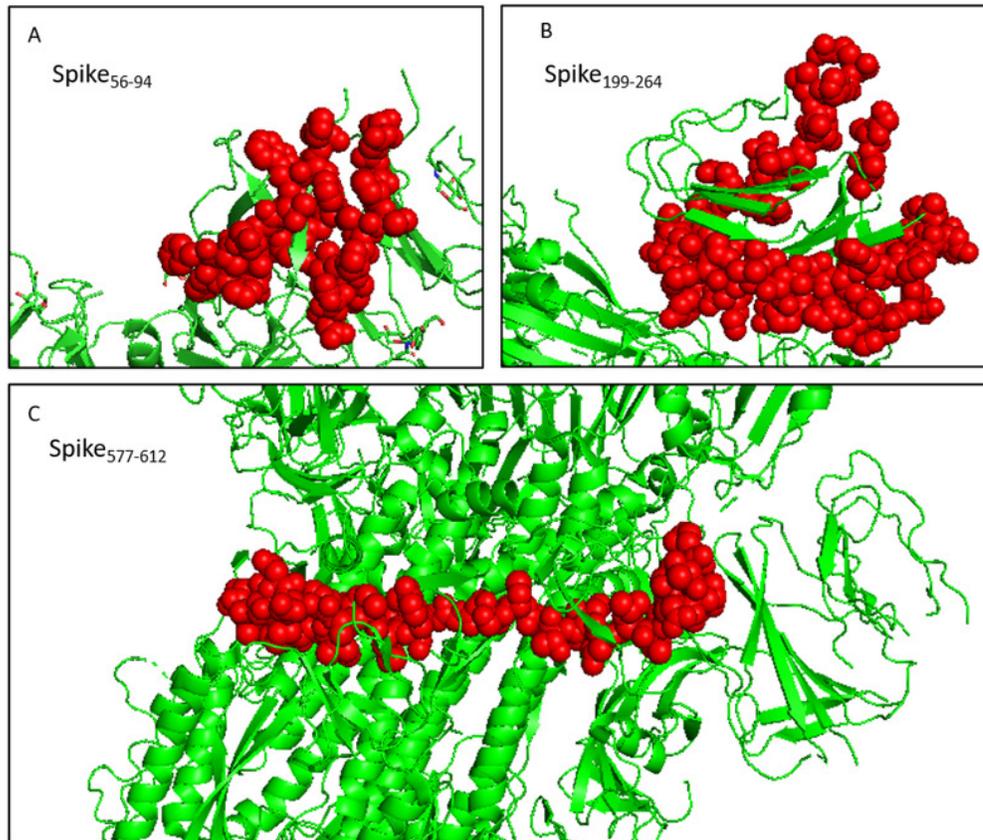


Fig. 6. Selected immunodominant fragments presented as spheres in the trimer structure of spike protein viewed by PyMOL

Selected fragments were presented as red spheres, green cartoons denote unselected sections. A, B, and C denote fragments Spike₅₆₋₉₄, Spike₁₉₉₋₂₆₄, and Spike₅₇₇₋₆₁₂ respectively.

Figure 7

A schematic diagram of recombinant peptide composed of selected fragments and a PADRE epitope.

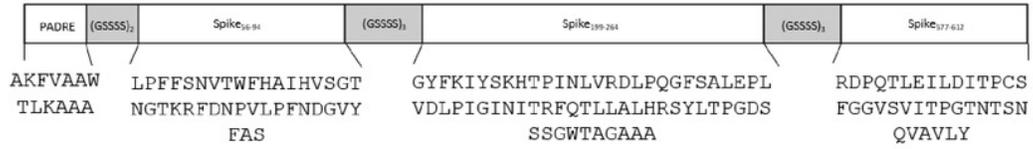


Fig. 7. A schematic diagram of recombinant peptide composed of selected fragments and a PADRE epitope.

Table 1 (on next page)

Linear B-cell epitopes predicted by ABCpred and BepiPred v2.0 with antigenicity score exceed the threshold value

1

Tools	Position	Sequence	Length	Antigenicity (cut off ≥ 0.4)
ABCpred	583-598	EILDITPCSFGGVSVI	16	1.3971
	406-421	EVRQIAPGQTGKIADY	16	1.3837
	415-430	TGKIADYNYKLPDDFT	16	0.9642
	648-663	GCLIGAEHVNNSEYCD	16	0.848
	288-303	AVDCALDPLSETKCTL	16	0.7905
	604-619	TSNQVAVLYQDVNCTE	16	0.7593
	307-322	TVEKGIYQTSNFRVQP	16	0.6733
	200-215	YFKIYSKHTPINLVRD	16	0.657
	257-272	GWTAGAAAYVGYLQP	16	0.621
	329-344	FPNITNLCPFGEVFNA	16	0.6058
	245-260	HRSYLTPGDSSSGWTA	16	0.6017
	280-295	NENGTITDAVDCALDP	16	0.5804
	49-64	HSTQDLFLPFFSNVTW	16	0.5305
	492-507	LQSYGFQPTNGVGYQP	16	0.5258
	70-85	VSGTNGTKRFDNPVLP	16	0.5162
	236-251	TRFQTLALHRSYLTP	16	0.5115
	266-281	YVGYLQPRTFLLKYNE	16	0.5108
	594-609	GVSVITPGTNTSNQVA	16	0.4651
320-335	VQPTESIVRFPNITNL	16	0.4454	
Bepipred v2.0	179-190	LEGKQGNFKNLR	12	1.1188
	404-426	GDEVQRQIAPGQTGKIADYNYKLP	23	1.1017
	14-34	QCVNLTTRTQLPPAYTNSFTR	21	0.7594
	56-81	LPFFSNVTWFHAIHVSGTNGTKRFDN	26	0.6041
	208-222	TPINLVRDLPPQGFSA	15	0.5531
	141-160	LGVYYHKNNKSWMESEFRVY	20	0.5308
	249-261	LTPGDSSSGWTAG	13	0.495
	306-321	FTVEKGIYQTSNFRVQ	16	0.4361
615-644	VNCTEVPVAIHADQLTPTWRVYSTGSNVFQ	30	0.4259	

Table 2 (on next page)

Details of epitopes in the preliminary immunodominant fragments selected according to linear B-cell epitope prediction results.

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Regions	Epitope predicted by ABCpred			Epitope predicted by Bepipred v2.0		
	Position	Sequence	Antigenicity	Position	Sequence	Antigenicity
49-85	49-64	HSTQDLFLPFFSNVTW	0.5305	56-81	LPFFSNVTWFHAIHV	0.6041
	70-85	VSGTNGTKRFDNPVLP	0.5162		SGTNGTKRFDN	
200-344	200-215	YFKIYSKHTPINLVRD	0.6570	208-222	TPINLVRDLPQGFSA	0.5531
	236-251	TRFQTLALHRSYLTP	0.5115	249-261	LTPGDSSSGWTAG	0.4950
	245-260	HRSYLTPGDSSSGWTA	0.6017			
	257-272	GWTAGAAAYVGYLQP	0.6210			
	266-281	YVGYLQPRTFLKYNE	0.5108			
	280-295	NENGTITDAVDCALDP	0.5804	306-321	FTVEKGIYQTSNFRV Q	0.4361
288-303	AVDCALDPLSETKCTL	0.7905				
307-322	TVEKGIYQTSNFRVQP	0.6733				
320-335	VQPTESIVRFPNITNL	0.4454				
329-344	FPNITNLCPFGEVFNA	0.6058				
583-663	415-430	TGKIADYNYKLPDDFT	0.9642	615-644	VNCTEVPVAIHADQL TPTWRVYSTGSNVFQ	0.4259
	583-598	EILDITPCSFGGVSVI	1.3971			
	594-609	GVSIVITPGTNTSNQVA	0.4651			
	604-619	TSNQVAVLYQDVNCTE	0.7593			
	648-663	GCLIGAEHVNNSYECD	0.8480			

Table 3 (on next page)

MHC-II and MHC-I binding epitopes predicted by TepiTool server with antigenicity score exceed threshold value

Type	Position	Sequence	Length	Allele	Core (smm-align)	Core (nn-align)	Percentile Rank	Antigenicity (cut off ≥ 0.4)
MHC-II binding	538-552	CVNFNFNGLTGTGVL	15	H2-IAb	FNFNGLTGT	FNFNGLTGT	8.55	1.3281
	374-388	FSTFKCYGVSPKLN	15	H2-IAb	FKCYGVSP	YGVSPKLN	6.45	1.0042
	199-213	GYFKIYSKHTPINLV	15	H2-Iab	KIYSKHTPI	YSKHTPINL	6.9	0.9278
	18-32	LTRTQLPYPAYNSF	15	H2-IAb	TRTQLPPAY	TRTQLPPAY	9.9	0.79
	60-74	SNVTWFHAIHVSGTN	15	H2-IAb	VTWFHAIHV	TWFHAIHVS	9.1	0.7044
	263-277	AAYYVGYLQPRTFLL	15	H2-IAb	VGYLQPRTF	VGYLQPRTF	8.75	0.6073
	592-606	FGGVSVITPGTNTSN	15	H2-IAb	VITPGTNTS	VSVITPGTN	6	0.5825
	238-252	FQTLALHRSYLTTPG	15	H2-IEd	TLLALHRSY	TLLALHRSY	9.85	0.5789
	345-359	TRFASVYAWNRKRIS	15	H2-IAb	FASVYAWNR	YAWNRKRIS	7.45	0.4963
	215-229	DLPQGFSALEPLVDL	15	H2-IAb	FSALEPLVD	FSALEPLVD	6.05	0.4812
	140-154	FLGVYYHKNNKSWME	15	H2-IEd	GVYYHKNNK	YYHKNNKSW	6.4	0.4793
	512-526	VLSFELLHAPATVCG	15	H2-IAb	FELLHAPAT	FELLHAPAT	2.9	0.4784
	87-101	NDGVYFASTEKSNII	15	H2-Iab	YFASTEKSN	VYFASTEKS	6.85	0.4277
	52-66	QDLFLPFFSNVTWFH	15	H2-IAb	FLPFFSNVT	FLPFFSNVT	2.95	0.4159
233-247	INITRFQTLALHRS	15	H2-IAA	ITRFQTLA	ITRFQTLA	1.9	0.4118	
MHC-I binding	643-651	FQTRAGCLI	9	H-2-Kk			0.6	1.7332
	612-620	YQDVNCTEV	9	H-2-Db			0.4	1.6172
	539-547	VNFNFNGLT	9	H-2-Kb			0.47	1.5069
	503-511	VGYPYRVV	9	H-2-Kb			0.47	1.4383
	379-387	CYGVSPKTL	9	H-2-Kd			0.3	1.4263
	16-24	VNLTRTQL	9	H-2-Kb			0.86	1.3468
	510-518	VVLSFELL	9	H-2-Kb			0.43	1.0909
	202-210	KIYSKHTPI	9	H-2-Kb			0.27	0.7455
	168-176	FEYVSQPFL	9	H-2-Kk			0.5	0.6324
	268-276	GYLQPRTF	9	H-2-Kd			0.2	0.6082
	505-513	YQPYRVVVL	9	H-2-Dd			0.3	0.5964
	488-496	CYFPLQSYG	9	H-2-Kd			0.64	0.578
	215-223	DLPQGFSA	9	H-2-Dd			0.69	0.5622
	342-350	FNATRFASV	9	H-2-Kb			0.56	0.5609
	84-92	LPFNDGVYF	9	H-2-Ld			0.21	0.5593
	484-492	EGFNCFYFPL	9	H-2-Kb			0.84	0.5453
	62-70	VTWFHAIHV	9	H-2-Kb			0.61	0.5426
	489-497	YFPLQSYGF	9	H-2-Dd			0.8	0.5107
	350-358	VYAWNRKRI	9	H-2-Kd			0.7	0.5003
60-68	SNVTWFHAI	9	H-2-Kb			0.82	0.4892	
262-270	AAAYYVGYL	9	H-2-Kb			0.98	0.4605	

Table 4(on next page)

Details of candidate immunodominant fragments adjusted according to the MHC-II binding T-cell epitopes prediction results.

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Regions	Linear B-cell epitopes				MHC-II binding epitopes		
	Tools	Position	Sequence	Antigenicity	Position	Sequence	Antigenicity
14-34	Bepipred v2.0	14-34	QCVNLTTRTQLPPAYTN SFTR	0.7594	18-32	LTTRTQLPPAYTNSF	0.7900
49-101	Bepipred v2.0	56-81	LPFFSNVTWFHAIHVSG TNGTKRFDN	0.6041	52-66	QDLFLPFFSNVTWFH	0.4159
	ABCpred	49-64	HSTQDLFLPFFSNVTW	0.5305	60-74	SNVTWFHAIHVSGTN	0.7044
	ABCpred	70-85	VSGTNGTKRFDNPVLP	0.5162	87-101	NDGVYFASTEKSNII	0.4277
199-359	Bepipred v2.0	208-222	TPINLVRDLPQGFSA	0.5531	199-213	GYFKIYSKHTPINLV	0.9278
		249-261	LTPGDSSSGWTAG	0.4950			
		306-321	FTVEKGIYQTSNFRVQ	0.4361			
	ABCpred	200-215	YFKIYSKHTPINLVRD	0.6570	215-229	DLPQGFSALEPLVDL	0.4812
		236-251	TRFQTLALHRSYLTP	0.5115			
		245-260	HRSYLTPGDSSSGWTA	0.6017			
		257-272	GWTAGAAAYVGYLQP	0.6210	233-247	INITRFQTLALHRS	0.4118
		266-281	YVGYLQPRTFLLKYNE	0.5108	238-252	FQTLALHRSYLTPG	0.5789
		280-295	NENGTITDAVDCALDP	0.5804			
		288-303	AVDCALDPLSETKCTL	0.7905			
		307-322	TVEKGIYQTSNFRVQP	0.6733	263-277	AAYYVGYLQPRTFLL	0.6073
		320-335	VQPTESIVRFPNITNL	0.4454			
329-344	FPNITNLCPFGEVFNA	0.6058					
345-359	TRFASVYAWNKRKIS	0.4963					
583-620	ABCpred	583-598	EILDITPCSFGGVSVI	1.3971	592-606	FGGVSVITPGTNTSN	0.5825
		594-609	GVSVITPGTNTSNQVA	0.4651			
		604-619	TSNQVAVLYQDVNCTE	0.7593			

Table 5 (on next page)

Significant features of the selected immunodominant fragments. The sequences marked with gray shading in the table represent amino acids with hydrophilicity and surface accessibility respectively.

Fragments	Spike ₅₆₋₉₄	Spike ₁₉₉₋₂₆₄	Spike ₅₇₇₋₆₁₂
Length(aa)	39	66	36
Sequence	LPFFSNVTWFHAIHVS GTNGTKRFDNPVLPF NDGVYFAS	GYFKIYSKHTPINLVRDLPQGFSALEPLVD LPIGINITRFQTLALHRSYLTPGDSSSGW TAGAAA	RDPQTLEILDITPCSFG GVSVITPGTNTSNQVA VLY
Antigenicity	0.4590	0.5774	0.9127
Domain	S1(NTD)	S1(NTD)	S1
Hydrophilicity fragments	LPFFSNVTWFHAIHVS GTNGTKRFDNPVLPF NDGVYFAS	GYFKIYSKHTPINLVRDLPQGFSALEPLVD LPIGINITRFQTLALHRSYLTPGDSSSGW TAGAAA	RDPQTLEILDITPCSFG GVSVITPGTNTSNQVA VLY
Surface Accessibility fragments	LPFFSNVTWFHAIHVS GTNGTKRFDNPVLPF NDGVYFAS	GYFKIYSKHTPINLVRDLPQGFSALEPLVD LPIGINITRFQTLALHRSYLTPGDSSSGW TAGAAA	RDPQTLEILDITPCSFG GVSVITPGTNTSNQVA VLY
Toxicity	Non-toxin	Non-toxin	Non-toxin
Allergenicity	non-allergen	non-allergen	probable allergen

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Table 6 (on next page)

The structure and antigenicity of final recombinant peptides

Final construct	PAN DR + (GGGS) ₂ + Spike ₅₆₋₉₄ + (GGGS) ₃ + Spike ₁₉₉₋₂₆₄ + (GGGS) ₃ + Spike ₅₇₇₋₆₁₂
Sequence	AKFVAAWTLKAAAGGGGSGGGSLPFFSNVTWFHAIHVSGTNGTKRFDNPVLPFNDGVYFASGGGSGGGGS GGGSGYFKIYSKHTPINLVRDLPQGFSALEPLVDLPIGINITRFQTLALHRSYLTPGDSSSGWTAGAAAG GGGSGGGGSGGGSRDPQTLEILDITPCSFGGVSVITPGTNTSNQVAVLY
Antigenicity	0.5690

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