**Designing a broad-spectrum four-helix bundle targeting different**

**strains of SARS-CoV-2 Spike Receptor Binding Domains with ACE2-**

**like binding interface**

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## Abstract

30 One major strategy for the severe acute respiratory syndrome coronavirus 2

31 (SARS-CoV-2) to evade antibody drugs or preventive vaccines is high mutation rate of

32 the spike receptor binding domain (RBD). Because variable RBDs of different SARS-

33 CoV-2 strains must bind to the same human receptor angiotensin-converting enzyme 2

34 (hACE2) for viral cell entry and infection, we hypothesize that designing a protein with

the same or very similar hACE2 binding interface might have a broad-spectrum effect against various SARS-CoV-2 strains. The designed protein binds specifically to the WT-RBD (with micromolar affinity) but not to RBDs from other SARS-CoV-2 strains. However, two rounds of the *E. coli* display and Magnetic Cell Sorting （MACS） selection are sufficient to yield a protein named CYN1 with nanomolar binding affinities not only to the WT-RBD but also to those of Omicron BA.1, XBB.1.16, and JN.1. Molecular dynamics simulations and free-energy hotspot analysis revealed that CYN1's broader spectrum capability stems from its engagement of essentially all ACE2 hotspot residues critical for WT-RBD binding, unlike the designed protein. The discovery of CYN1, differing by only four mutations from the designed protein, confirms that targeting the small interface of human viral receptors—rather than the entire receptor—offers a viable strategy for developing broad-spectrum inhibitors. This approach minimizes potential off-target effects arising from receptor multifunctionality.

## Introduction

Severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) quickly became a worldwide pandemic after it was first identified in December 2019. As of December 22th 2024, this virus has infected nearly 1 billion people and caused over seven million deaths according to the World Health Organization (COVID-19 cases | WHO COVID-19 dashboard, https://data.who.int/dashboards/covid19/cases?n=c). The SARS-CoV-2 virus enters the host cell through the binding of its trimeric spike protein with Angiotensin Converting Enzyme 2 (ACE2) on the human cell surface. The receptor-binding domain (RBD) on the S1 subunit of the spike protein is responsible for binding to ACE2 [1]. The main strategy for the virus to escape the human's immunity (either from previous infection or from the vaccines [2–6]) is evolved mutation on the spike RBD. Many broad-spectrum antibodies and vaccines were proposed and tested but yet to be successful [7,8].

While the spike RBD constantly mutates, our human receptor hACE2 does not change. Thus, it is conceivable that employing hACE2 to bind with the spike RBD might offer broad-spectrum binders of spike RBDs from different virus strains. Indeed, the engineered hACE2 [9–11] were proposed with a demonstrated broad-spectrum effect. However, hACE2, as a hydrophobic transmembrane protein, relied on a solubilizing label [9–13] to be effective. Moreover, ACE2 has many functions [14] and direct use of ACE2

may have unintended side effects. De novo designed proteins/peptides for blocking the

binding of spike to ACE2 were also proposed [15–17], but they are not expected to have

the broad-spectrum capability of ACE2 as they have significantly modified ACE2.

Here, we designed a four-helix bundle protein based on the short 22-residue

double-helical region on ACE2 that interacts with spike's RBD. We subsequently

improved it by directed evolution with a demonstrated strong binding to several spike's

RBD variants. The mechanism of this broad-spectrum capability was elucidated by

conducting molecular dynamics simulations and binding free energy analysis by

combining enthalpy and entropy components derived from MM/GBSA[18–20] and

interaction entropy (IE) methods[21] with computational Alanine scanning (AS)

(ASGBIE method[22–25]) for hotspot analysis[26–29]. The result confirms that improved

capture of hotspot residues of ACE2 by the four-helix bundle through directed

evolution is the reason behind its improved broad-spectrum capability.

## Materials and Methods

### SPalign

83 SPalign [30] is a structural alignment program developed by us that compares two protein

84 structures using a size-independent scoring function SP-score. Here it was employed to

85 locate the best natural backbone structure for a manually constructed four-helix bundle

86 (H4) based on two helices in ACE2 located at the hACE2-SARS-CoV-2 RBD interface.

### OSCAR-loop

88 OSCAR-loop[31] is a program for building loops for a protein region with missing amino

89 acid residues. OSCAR-loop was kindly provided by Dr. Shide Liang. Here the program

90 was employed to build the loop regions for the four-helix bundle H4.

### OSCAR-Design

OSCAR-Design[32] is a protein design program that was developed by us to design

protein sequences for a given protein backbone. This program employed an energy

function that was based on series expansion with coefficients optimized for protein

design. Here the program was employed to design protein sequences given a backbone

structure and fixed sidechain conformations of RBD-binding interfacial residues copied

from hACE2.

**DbD2**

DbD2 (Disulfide by Design) [33] is a web-based program for designing disulfide bonds

in proteins, which is provided at   http://cptweb.cpt.wayne.edu/DbD2/. Here the

program was used to add a disulfide bond to improve the stability of a designed protein

as a monomer.

**Plasmids**

**Protein expressions**: The H4 sequences designed and a positive control AHB2[15] (A

reported RBD binder) were optimized for *Escherichia coli* (*E. Coli*) expression system

and directly synthesized and subcloned into pET-21a (+) vector (BGI Genomics,

Beijing).

***E. coli* display**: error prone-PCR (EP-PCR) was used to introduce mutations into a

target sequence using QuickMutation™ Random Mutagenesis Kit (D0219S, Beyotime)

with about 0.8% mutation rate. The mutated DNA fragments were subcloned into *E.coli*

display vector pDSG323-Null-Myc (115594, Addgene) between the *Kpn*I and *Bam*HI

restriction sites for construction of the mutation library.

**Gene knockout:** In order to reduce the interference of fimbriate on *E. coli* display, we

knocked out the *fimA-H* from MG1655 strain using pEcCas/pEcgRNA system[34].

pEcCas (73227, Addgene) and pEcgRNA (166581, Addgen) were purchased from

Addgene. The specific sgRNA sequence was designed using CRISPOR[35] with MG1655

genetic background and then constructed into the pEcgRNA vector, named as

pEcgRNA-1. The homology arm fragments (sequence in Table S6) were directly

synthesized as donor DNA and subcloned into between the EcoRV and PciI restriction

sites of pUC57-1.8k vector (General Biol, China).

All amino acid sequences and DNA sequences were listed in Tables S5 and S6,

respectively.

**Construction of MG1655$^{\Delta fimA-H}$ and Generation of *E.coli* display libraries**

The modified CRISPR/Cas9 system (pEcCas/pEcgRNA) [34] was used to knock out the

*fimA-H* gene for construction of the MG1655$^{\Delta fimA-H}$ strain. The pEcCas was first

transformed into the MG1655 strain. Then the MG1655-pEcCas strain was co-

transformed with the pEcgRNA plasmid and donor DNA by electroporation, and

recovered for 1 h in 2 mL LB at 37 °C. The clones were randomly picked for colony

PCR and Sanger sequencing for verifying the knockout results. Correct colonies were

subjected to the plasmid curing process to obtain empty MG1655$^{\Delta fimA-H}$ strains as

described in the reference[34].

The constructed library plasmid was transformed into electrocompetent

MG1655$^{\Delta fimA-H}$ strain by electroporation. The size of each library was $1 \times 10^6$ clones, as

determined by plating on LB-Cm agar plates with 2% w/v glucose incubated at 30 °C.

***E. coli* display and Magnetic Cell Sorting (MACS)**

We followed the previously reported *E. coli* display protocol[36]. More specifically, the

*E. coli* library was induced by 1 µg/mL anhydrotetracycline (ATc) at 37 °C for

overnight and harvested by centrifugation (3000 × g, 3 min). It is then washed five

times with 2 mL PBS and resuspended in 1 mL of PBS. Afterwards, 100 µL of the

bacteria was mixed with 50 nM Biotinylated RBD and adjusted to the final volume of

200 µL with PBS-BSA (PBS supplemented with 0.5% w/v BSA, sterile filtered and

degassed), which was incubated at room temperature for 30 min. Thereafter, the

bacteria were washed five times with 1 mL of PBS-BSA, resuspended in 100 µL of the

same buffer containing 20 µL of anti-biotin paramagnetic beads (Miltenyi Biotec), and

incubated at 4 °C for 20 min. Next, the bacteria were washed three times with 1 mL of

PBS-BSA and resuspended in 500 µL of the same buffer, of which 10 µL was kept

aside to calculate the input bacteria before the MACS procedure and the remaining

portion (490 µL) was applied onto a MACS MS column (Miltenyi Biotec), that was

previously equilibrated with 500 µL of PBS-BSA and placed on the OctoMACS

Separator (Miltenyi Biotec).

The unbound cells (flow-through) were collected, and the column was washed five

times with 2 mL of PBS-BSA. The wash was combined with the flow-through as the

"Unbound fraction". Next, the column was removed from the OctoMACS Separator

and placed onto a new collection tube to free the bound fraction by washing with 2 mL

of PBS-BSA. The bound fraction was applied onto a new equilibrated MACS MS

column for reduced bacterial non-specific binding. The column was washed five times

with 2 mL of PBS-BSA[36]. Next, the low-affinity mutants were washed away by 10 mM

Glycine-HCl pH 2.0 and pH 3.0 (two times with 2 mL each time, and three times for

second-round evolution), and the high-affinity mutants were still on the column.

Afterwards, the column was removed from the OctoMACS Separator and placed onto

a new collection tube, 2 mL of LB was added, and the cells were eluted out. This

fraction was labeled as the "Screen fraction". Serial dilutions of screen fractions were

plated to determine CFU and perform the Sanger sequencing.

**Expression and purification of recombinant proteins**

Expression plasmids were transformed into chemically competent *E. coli* Shuffle T7-

B. Protein expression was then induced with 0.1 mM of isopropyl β-D-

thiogalactopyranoside (IPTG) at 16 ℃. After overnight induction, the cells were

harvested by centrifugation at $10,000 \times g$ for 30 min and then resuspended in lysis

buffer (150 mM NaCl, 20 mM Tris-HCl (pH 8.0), with 0.1 % triton X-100). The cells

were lysed with an Ultrasonic Homogenizer (JY92-IIN, Scientz) for 20 minutes (2 sec

on-3 sec off) at an amplitude of 30 %, and then centrifuged at $30,000 \times g$ for 30 min.

The precipitate was then removed, and the supernatant was collected. The supernatant

was purified by affinity chromatography (HisTrap FF crude 5 mL, Cytiva) and followed

by size exclusion chromatography (HiLoad 16/600 Superdex 75 pg, or Superdex 75

Increase 10/300 GL, Cytiva) on the ÄKTA pure™ 25M system (Cytiva). All protein

samples were characterized using 4-20 % gradient SDS-PAGE and their purity was

higher than 95%. Protein concentrations were determined using a NanoDrop

spectrophotometer (Thermo Scientific) with absorbance at 280 nm using predicted

extinction coefficients (https://web.expasy.org/protparam/).

**Circular dichroism spectroscopy**

Far-ultraviolet CD measurements were carried out with a JASCO-1500 equipped with a temperature-controlled multi-cell holder. Wavelength scans were recorded from a wavelength region of 198 - 255 nm at 25 ℃, 95 ℃, and 25 ℃ after melt, and taking the average of four repeats. Temperature melts monitored dichroism signal at 222 nm in steps of 2 ℃/minute from 16 to 95 ℃ with 30 s of equilibration time [15]. Wavelength scans and temperature melts were performed using 0.025 mg/mL protein in PBS buffer (1.8 mM $KH_2PO_4$, 10 mM $Na_2HPO_4$, 137 mM NaCl, 2.7 mM KCl, pH 7.4) with a 1 mm path-length cuvette.

Melting temperatures were determined by fitting the data with a sigmoid curve equation by the instrument-integrated software. Helical contents were analyzed according to the reported method[37]: negative ellipticity at 208 and 222 nm of the CD spectrum. The α-helical content was predicted using K2D3 web server (K2D3: Estimating protein secondary structure from CD spectra) [38] with the following parameters: (1) the CD spectra from 198 to 240 nm was converted to molar residue ellipticity (MRE, deg·cm²·dmol$^{-1}$) units prior to analysis; (2) structural prediction constraints include specifying protein length as 146 amino acid residues.

**Biolayer interferometry**

Biolayer interferometry (BLI) binding data were collected in Octet RED384 (ForteBio) at 30 ℃ and processed using the instrument-integrated software. All RBDs were biotinylated by the biotinylating kit (GENEMORE, Cat. G-MM-IGT) according to the kit's manual. The biotinylated RBDs were loaded onto streptavidin-coated biosensors (SA ForteBio, Cat. 18-5019) in PBS (pH 7.4) buffer. The buffer of analyte proteins was exchanged to binding buffer (PBS, 0.05% surfactant P20) by Amicon® Ultra Centrifugal Filter (Millipore) and then diluted to the desired concentrations. After baseline measurement with the binding buffer alone, the binding kinetics were monitored by dipping the biosensors in wells containing the target protein at the indicated concentration (association step) and then dipping the sensors back into baseline/buffer (dissociation). The data were processed using ForteBio Data Analysis

215 software 12.0, and the equilibrium dissociation constant ($K_D$) values were obtained by

216 global fitting with 1:2 and 1:1 binding models.

**SARS-CoV 2 RBDs**

The RBDs of SARS-CoV 2 variants (WT, XBB.1.16 and JN.1) were purchased from

Sino Biological (Cat. 40592-V08B, 40592-V08H136, 40592-V08H155), and the RBD

of BA.1 was kindly provided by Haitao Yang's Group. We chose the Omicron BA.1,

XBB.1.16 and JN.1 as the representatives of virial strains for testing our designed

proteins because these strains were a part of the important SARS-CoV-2 evolution

nodes, and have a high immune evasion ability to the anti-spike antibodies or vaccines

that were developed based on their strain's ancestors [7].

**Molecular Dynamics Simulations**

The complex structure of ACE2 and WT SARS-CoV-2 Spike RBD (WT-RBD) was

obtained from PDB (PDB ID: 6M0J). As each designed protein has two binding

interface to WT-RBD, the complex structures of WT-RBD with D2-133 in either its

helices 1-2 or helices 3-4 regions were predicted by OSCAR-Design (Figure 1). Other

complex structures of CYN1, E6, and E9 in their helices 1-2 or helices 3-4 regions with

WT-RBD were mutated from D2-133. All complexes were filled by the leap module in

AMBER20 package[39]. All MD simulations were performed using pmemd.cuda in

AMBER20 with ff14SB force field for the protein in a truncated octahedron box formed

by TIP3P explicit water molecules. The closest distance between any atoms originally

present in solute and the edge of the periodic box was 12 Å. The particle mesh Ewald

(PME) was used to treat the long-range electrostatic interactions. The nonbonded

interactions were truncated with 10 Å cutoff. Periodic boundary condition (PBC) was

imposed on the system during the calculation of nonbonded interactions. The time-step

was set at 2 fs and SHAKE[40] was used to constrain the bonds involving hydrogen atom.

Langevin thermostat with the collision frequency of 2.0 was applied to control the

temperature. Firstly, the system was minimized with the protein constrained to

equilibrate the solvent. Then, the protein was released to minimize the whole simulation

system. Afterward, the system was slowly heated to 300 K, followed by a 40 ns

equilibration of the whole system in an NPT ensemble at an interval of every 10 fs, and

during this session, no restrain was performed. Then, a 10ns production run was

performed to obtain 5000 snapshots for free-energy and hotspot analysis by ASGBIE.

**Free-energy and hotspot analysis by ASGBIE**

The ASGBIE method[22–25]combines enthalpy and entropy components derived from

MM/GBSA[18–20] and interaction entropy (IE) methods[21] with computational Alanine

scanning (AS) for hotspot analysis[26–29]. Here, we simply followed the same method

utilized in Ref. [22–25].

**Results**

**Figure 1. The flowchart of the iterative design and improvement of a four-helix**
**bundle (H4).** The designed protein has two copies of the two-helix binding interface
of hACE2 (helices 1-2 and 3-4) to SARS-CoV-2 Spike receptor binding domain (RBD),
whose backbone structure was modeled after the closest four-helix template in the
protein databank. The sequences for this backbone structure along with fixed binding
residues were designed by OSCAR-Design and five were tested for binding affinity. To
improve monomeric stability, the second-round design was performed after adding a
disulfide bond. The best design (D2-133) was improved by directed evolution using *E.*
*coli* display and Magnetic Cell Sorting (MACS) to obtain the final protein CYN1.

The flowchart of the overall design, subsequent screening and redesign as well as final

improvement of four-helix ACE2-interface mimics by directed evolution is shown in

Figure 3. Specific details are described as below.

**Initial design of a four-helix bundle (H4) with two ACE2-Spike binding interfaces**

**Backbone Structure**: We obtained the interface residues between human ACE2

(hACE2) and wild-type SARS-CoV-2 Spike protein according to their complex

structure (PDB 6M0J) based on contact distances and the reduction of solvent

accessible surface area (SASA) per residue after complex formation. Most interface

residues in hACE2 were found located on the two N-terminal helices. After careful

inspection of the conformation of the two helices, we built a four-helix bundle prototype

by manually dimerizing the two N-term helices of hACE2. This manually built dimeric

structure together with binding Spike proteins was further relaxed using ROSETTA[41].

The relaxed structure was then compared to all existing four-helix bundle structures in

RCSB PDB with SP-align[30]. All structurally aligned PDBs were sorted according to the

SPe-alignment score and top 10 ranked structures were manually inspected and the

structure of the serine-rich domain from Crk-associated substrate (PDB 1Z23) was

found as the closest match and employed to adjust the backbone structure of the

prototype by modifying the relative distance and orientation between dimerized two-

helix bundles.

**Sequence Design**: We copied the conformations of ACE2 interface residues onto

the prototype of the four-helical bundle: residue 22-84 from ACE2 in 6M0J were used

to build the N-term half of the four-helix bundle, with P84 modified to G84, and

residues 22-88 from ACE2 in 6M0J were extracted to build the C-term half. Residues

85-88 were included in building the C-term half, because they mimicked well the

terminal cap of the template H4 in PDB 1Z23, which may contribute to the stability of

designed H4. The gap between the two halves was connected by building the loops of

various lengths (5-9) modeled using OSCAR-loop [31]. Note that residue 84 from N-term

and residue 22 from C-term were freely mutated during loop modeling and were

included in loop modelling. All residues were then re-numbered starting from 1. The

H4 structure was built to have two binding interfaces to interact with the Spike protein.

The above constructed structures with different loop lengths were further

optimized with AmberTools [42] and inspected with Procheck [43]. One backbone structure,

which has a modeled loop of 9 residues passed the check. Thus, this backbone structure

was employed to redesign by OSCAR-Design [32] with 2x22 fixed interface residues (N-

term half: E2, Q3, T6, F7, D9, K10, H13, E14, E16, D17, Y20, Q21, L24, N40, N43,

K47, F51, E54, Q55, L58, M61 and Y62; C-term half:    E72, Q73, T76, F77, D79, K80,

H83, E84, E86, D87, Y90, Q91, L94, N110, N113, K117, F121, E124, Q125, L128,

M131 and Y132) in the presence of two fixed Spike proteins (Figure 1).

**Sequence Selection**: Five thousand sequences designed by OSCAR-Design were

clustered with CD-HIT [44] and filtered according to the area of the largest hydrophobic

surface-patch $S_{HPA}$, the number of low-complexity areas $N_{LCA}$ and the number of

atomic clashes $N_{clash}$. The areas of hydrophobic surface-patches were calculated by

QUILT [45]. The number of low-complexity areas was calculated by segmasker from the

NCBI BLAST+ package [46]. We found that there were five sequences with sequence

identity lower than 75% between each other, $S_{HPA} \leq 550Å^2$ (509 $Å^2$ for the original

sequence), $N_{LCA} = 0$ and $N_{clash} = 0$. These sequences (D0064, D0420, D0753,

D4937, and D4983) were selected for the first-round experimental test (Table S1).

Three of the five designs (D0753, D4937, and D4983) were expressed soluble and

purified (Figure S1). All have micromolar binding affinities to WT-RBD with D4983

having the strongest binding affinity (Figure S2). However, these three designed

proteins all existed in aggregated forms on a chromatographic column (Figure S3).

**Adding Disulfide-bond and Redesigning Sequences**: To make the monomeric

folding of the designed proteins more stable, we introduced one disulfide bond to the

backbone structure in its non-interfacial areas by using the program DbD2 (Disulfide

by Design) [33]. This process produced several designs with the disulfide bond added to

various locations. These designs were further structurally optimized and sequence-

redesigned in the same way as in the first-round design with the disulfide-bond fixed.

We collected 1000 designs and further filtered with $S_{HPA} \leq 385Å2$, $N_{LCA} = 0$, and

$N_{clash} = 0$. This results in six second-round sequences with disulfide bonds located at

CYS1-CYS133  (D2-133), CYS22-CYS48  (D2-128), CYS25-CYS44  (D2-477),

CYS70-CYS136 (D2-247), CYS92-CYS118 (D2-830) and CYS95-CYS114 (D2-099),

respectively (Figure 1 and Supplementary Table S2). Among 6 new designs (D2-099,

D2-128, D2-133, D2-247, D2-477, D2-830), only D2-133 was monomeric, D2-099,

D2-477 were partly monomeric, whereas D2-128, D2-247, and D2-830 were mostly

multimeric (Figure 2A, 2B and 2C). Thus, only D2-133 was employed for the binding

affinity test.

**Figure 2. Expression and purification of the second-round designs of H4 along
with affinity measurement**. Size-exclusion chromatography for D2-133 (**A**), D2-099
and D2-477 (**B**), and D2-128, D2-247, and D2-830 (**C**) indicates that only the D2-133
design is purely monomeric. (**D**) Purified D2-133 was shown on a 4-20% SDS-
polyacrylamide gel and stained with Coomassie brilliant blue. The binding affinity
between D2-133 and WT SARS-CoV-2 Spike RBD at the micromolar level according
to the Biolayer-interferometry (BLI) measurement (Fitted by 1:1 model (**E**) and 1:2
model (**F**)).

**The designed ACE2 mimic D2-133 exhibited micromolar affinity against RBD
domain of Spike protein**

The purified monomeric D2-133 was obtained from Size-exclusion chromatography

(Figure 2D). Its binding affinity to RBD of wild type Spike was determined using BLI

(Figure 2E and 2F). The D2-233 showed a micromolar affinity (1.3 μM). This affinity

was similar to the AHB2 (affinities ranging from 100 nM to 2 μM) [15] or DBR3_02 ($K_D$:

4.6 μM), previously designed RBD binders [17]. Compared with the 1:1 fitting model

(Figure 2E), D2-133 showed a better fitting with the 1:2 fitting model (Figure 2F),

consistent with our inclusion of two hACE2-RBD binding interfaces from hACE2.

**Improving D2-133 by *E. coli* display and MACS selection.**

The *E. coli* display was used to improve the D2-133 affinity. To exclude the influence

of fimbriae on display[36], *fimA-H* was knocked out from MG1655 using

pEcCas/pEcgRNA system [34] and the MG1655$^{\Delta fimA-H}$ strain was successfully constructed

for bacterial display (see Figure S4 and Methods). EP-PCR was used to introduce

mutations on the D2-133 sequence for the construction of *E. coli* display library with

about 0.8% mutation rate.

The D2-133 library can bind to RBD-labeled beads but not to empty beads as

shown in Figure 3A (Group 1 vs Group 2). However, neither strong acid (Figure 3A,

Group 3) nor high salt (Figure 3A, Group 4) conditions can remove low-affinity

mutants from the library. Thus, we increased the selection pressure by lowering the pH

in round 1 (Figure 3B, Group C and D). Two clones (labeled as E-6 and E-9) from

round 1 were selected for expression and purification for affinity evaluation using BLI.

Both E-6 and E-9 have double mutations over D2-133 (K70E and K90R for E-6 and

N103I and E125G for E-9). Both improve over D2-133 in binding affinity to WT-RBD

with $K_D$ of 99 nM and 501 nM, respectively (Figure 3B and 3C).

Employing the best-affinity sequence E-6, we performed the second round of *E.*

*coli* display and MACS selection (Figure 3A). One clone CYN1 was selected. It has

four mutations over D2-133 (L59T, K70E, K90R, L129T). CYN1 was expressed and

purified in the monomeric state (Figure 3D). It has a binding affinity of 22 nM to WT

Spike RBD, about 60 times stronger than that of D2-133 (Figure 2 F vs Figure 3 E) and

4 times stronger than that of hACE2 with WT Spike RBD [47–49].

**Figure 3. Improved binding affinity of D2-133 by *E. Coli* display and MACS. (A)**
Round 0 indicates poor selection of D2-133 mutants with pH 4.0 (Group 3) and 1M
NaCl (Group 4), compared to no selection pressure (Group 2). Increased selection
pressure at pH 2.0 and 3.0 in Round 1 led to E-6 and E-9. In Round 2, the selection of
E-6 mutants led to CYN1. (**B**) and (**C**) The binding affinities of E-6 and E-9 with WT-
RBD were measured by BLI. (**D**) Size-exclusion chromatography for CYN1. (**E**) The
binding affinity of CYN1 with WT-RBD measured by BLI.

**High thermal stability of CYN1**

To investigate the secondary structural content and thermal stability of D2-133 and
CYN1 at far-UV region, Circular dichroism (CD) experiments were carried out using
Circular Dichroism Spectrophotometer J-1500. The results of CD spectrum indicated
D2-133 and CYN1 showed typical characteristic peaks of the α-helical secondary
structure with negative ellipticity at 222 and 208 nm (Figure 4A and 4B), suggested

that their secondary structure was dominated by α-helix which was consistent with our

design of four-helix bundle. Although the CD intensity decreased as the temperature

increased from 25 ºC to 95 ºC, the negative peaks at 222 and 208 nm were retained,

indicating that the helical structure of the protein sample at room temperature retained

part of the secondary helical structure at high temperatures. Figure 4A and 4B further

showed the CD spectrum of the initial 25 ºC spectrum (black) compared with the

spectrum measured at 95 ºC (red) and at 25 ºC after the melt (green). The spectra before

and after refolding were very similar, indicating that the protein does refold once the

temperature reduced, confirming that the folding was reversible. The CD-spectrum-

derived helical contents according to the K2D3 web server is 52.2 % for D2-133 and

61.42 % for CYN1, suggesting that CYN1 has a more stable structural fold around the

designed helical conformation (Figure 4E, F). However, this is still lower than 91%

residues in helical regions for the designed protein.

**Figure 4. The Circular Dichroism Spectrum indicates high thermal and structural**
**stability of D2-133 and CYN1.** The Circular Dichroism Spectrum of D2-133 (**A**) and
CYN1 (**B**) at 95 ºC and 25 ºC at native and refolded conditions as labeled. The CD
signal of D2-133 (**C**) and CYN1 (**D**) at 222-nm wavelength, as a function of
temperature. The predicted secondary structure contents of D2-133 (**E**) and CYN1(**F**)
by K2D3 web server.

The CD signal strength as a function of temperature was monitored at 222 nm for
D2-133 and CYN1 (Figure 4C and 4D). However, there is a lack of significant
temperature-dependent changes suggesting that the melting temperatures of CYN1 and
D2-133 were both greater than 95 ºC. These results suggest that CYN1 exhibits the
same high thermal stability as the designed H4 protein D2-133.

**The CYN1 showed broad-spectrum binding capability against RBD mutants from three different virus strains**

To verify whether D2-133 and CYN1 bind broadly to RBDs of different SARS-CoV-2 strains, the RBDs of Omicron BA.1, XBB.1.16, and JN.1 virus strains were purchased for measuring binding affinity, and they had 13 mutations, 14 mutations, 27 mutations to the WT-RBD, respectively (Figure 5G). The results showed that CYN1 retains its activity in binding to RBDs from different virus strains (Figure 3E vs Figure 5A, 5B and 5C). Compared to the wild type RBD, the binding affinity of CYN1 to XBB.1.16 RBD was similar, at 17 nM, while the affinities of CYN1 to BA.1 and JN.1 RBDs were weakened, at 88 nM and 164 nM, respectively. The previously reported RBD binder AHB2 [15] was employed as a control, and its binding affinities to WT-RBD and BA.1 RBD were determined. As shown in Figure 5D, the binding affinity of AHB2 to WT-RBD was at 0.83 nM, consistent with the previous study [15]. However, the affinity of AHB2 to the Omicron BA.1 RBD was only at 815 nM (Figure 5E), reduced by 1000-fold compared to the wildtype RBD (Figure 5F). By comparison, only a 4-fold reduction of binding of CYN1 to BA.1 relative to the binding to WT was observed (Figure 3E). It should be emphasized that the original design D2-133 does not display significant binding with the RBDs of BA.1, XBB.1.16 and JN.1 (Figure S5), despite of only 4 mutations away from CYN1.

**Figure 5. Broad spectrum measurement of binding to RBDs of other SARS-CoV 2 strains.** Affinity of CYN1 with RBDs, fitted by 1:2 model to XBB.1.16 (**A**), BA.1 (**B**), JN.1 (**C**). Binding affinity of AHB2 fitted by 1:1 model to WT (**D**) and to BA.1 (**E**). (**F**) Table of Affinities of ACE2 [7,47,50,51], AHB2, and CYN1 with RBDs. (**G**) The mutation sites of the three Omicron RBDs to WT-RBD.

### Interpretation of binding free energy by molecular dynamics simulations and free energy calculations

To understand different behavior observed between D2-133 and CYN1, we performed molecular dynamics simulations of complexes of the WT-Spike RBD with ACE2, D2-133, E6, E9, and CYN1 and observed quite stable binding interfaces with the root mean square deviation (RMSD) values of the designed binding interfaces stable less than 3 Å for both 1-2 and 3-4 helices (Figures 6A and 6B). We then employed the ASGBIE method to calculate the interaction free energy between all residues within 10 Å of the binding interface and the other protein molecule. Fig. 6C shows a strong correlation with a Pearson correlation coefficient at 0.97 between experimental determined free energies and ASGBIE-calculated results according to simulated conformations. This consistency provides the basis for the following hot-spot analysis by computational alanine scanning.

**Figure 6.** (A) Root-mean-squared deviation of helices 1-2 interface from the
respective starting conformations by molecular dynamics simulations of ACE2, D2-
133, E6, CYN1, and E9 in complex with a single WT-RBD of 5000 snap shots of 10
ns production runs after 40 ns equilibration run. (B) as in (A) but for helices 3-4
interface. (C) Calculated versus experimentally measured binding free energies of
ACE2, D2-133, E6, CYN1, and E9 in complex with a single WT-RBD. (D)
Comparison of hotspot residues identified by computational alanine scanning in
helices 1-2 interface of D2-133 and CYN1 with the hotspot residues of ACE2 (E) As
in (D) but for helices 3-4 interface. The sequence number in ACE2 was changed to
match with those designed proteins for comparison.

**Hotspot residues and their implication on broad-spectrum capability**

For binding of ACE2 to WT-RBD, we found a total of 8 hotspot residues, ranked by
their contributions as follows: 20TYR, 13HIE, 10LYS, 16GLU, 62TYR, 3GLN, 2GLU,
and 6THR. These hotspot residues were included as a part of fixed interfacial residues
for designing ACE2 mimics. For our designed proteins, we have two surface regions
mimicking the ACE2 interface to WT-RBD (helices 1-2 and helices 3-4). Thus, we
performed two separated calculations to identify hotspots in two dual-helix regions. As
shown in Figures 6D-E and Supplementary Tables S3-4), D2-133 has 6 out of   8

hotspot residues in the helices 1-2 interface and 6 out of 7 hotspot residues in helices

3-4 interface overlapping with ACE2 hotspot residues. By comparison, CYN1 has more

hotspot residue interactions with WT-RBD with stronger binding affinities (10 for helix

1-2 and 11 for helix 3-4). Moreover, these hotspots overlap more with 8 ACE2 hotspot

residues (7 for helix 1-2 and 8 for helix 3-4). This improved recovery of ACE2 hotspot

residues by CYN1 explains its broader spectrum capability over D2-133.

**Discussion**

SARS-CoV-2 has a fast mutation rate because it is an RNA virus with billions of hosts

[52,53]. Much research shows that the spike's mutation was the most important factor for

SARS-CoV-2 to escape the vaccine protection developed by pharmaceutical companies

[3,4,6,7,54]. The spike's mutations also lead to resistance to antibody drugs because many

antibody drugs were based on antigens on the spike proteins that are subsequently

mutated [55,56]. Here, we examined the possibility of designing a broad-spectrum protein

by copying the same hACE2 interface that can bind to the RBDs of the different viral

strains.

Despite the simplicity of the above idea, the design progress is not as

straightforward as originally planned. In the first-round design, 2 out of 5 top-ranked

designs did not have soluble expression, despite we have chosen the sequences with the

lowest continuous hydrophobic patch area, indicating that the largest hydrophobic

patch area is not necessarily the best indicator for solubility. Nevertheless, three of the

five designs were expressed soluble and purified (Figure S1) and all bind with WT-

RBD with micromolar binding affinity (Figure S2). Although the binding affinity is

much lower than the binding between ACE2 and WT-RBD, the rate of success in

achieving binding by OSCAR-Design is high (60%).

However, these three expressed proteins existed in aggregated forms on the

chromatographic column (Figure S3). The difficulty of expression and aggregation

faced here is likely due to the high hydrophobic patch area of the ACE2 binding

interface and the need to maintain two large hydrophobic patches for our design with

two ACE2 binding interfaces. To address this problem, we incorporated a disulfide

bond in our design. All six designs are now expressed and soluble. Moreover, three of

the six designs tested now exist in significant monomeric form with D2-133 purely

monomeric, corresponding well to the fact that it has the smallest hydrophobic patch

area ($361\text{Å}^2$) and two with the largest hydrophobic patch areas ($382 \text{ Å}^2$ and $384 \text{ Å}^2$) are

all in multimeric form.

D2-133, however, continues to bind at the micromolar level. Moreover, it does not

bind to RBDs of other SARS-CoV-2 strains. This indicates that the D2-133 interface

grafted from ACE2 for RBD binding is only partially functional. That is, D2-133 must

not have all the original ACE2 interfacial residues in the right place to interact with

Spike RBD because it has a weak affinity to WT-RBD and no affinity to the RBDs of

other viral strains.

D2-133 was improved by a factor of 13 in binding affinity to WT-RBD with one

round of MACS selections. However, the only two mutations (K70E and K90R) in E6

are unexpectedly all outside the pre-defined ACE2-mimic interface region. The second

round of MACS selections added two more mutations. This time both mutations are in

the pre-defined ACE2 interfacial regions (L59T and L129T), but mutations are from

hydrophobic leucine to more hydrophilic threonine. One would expect the opposite as

less hydrophobic interface would lead to lower binding affinity.

To gain some understanding why these mutations yielded higher affinity and

improved resistance against RBD mutations, we conducted molecular dynamics

simulations on all systems to investigate the microstructural features of binding. Both

the 1-2 and 3-4 helical regions were simulated as binding sites, respectively. The

binding free energies of four-helix structures with WT-RBD obtained by using the

ASGBIE method have a near perfect correlation to experimental measured results. A

further hotspot analysis indicates that CYN1's hotspot residues (both helices 1-2 and

helices 2-4 regions) have more overlap with ACE2 hotspot residues than D2-133,

suggesting the likely reason behind improved broad-spectrum capability than D2-133.

CYN1 has not achieved a perfect broad-spectrum capability to constantly evolving

SARS-CoV-2. This is because the binding affinity of CYN1 to other strains' RBDs are

lower than those of hACE2 [7,57,58]. Moreover, there are many strains' RBDs we have not

tested yet due to the huge experiment cost. We had tried to determine the RBD-CYN1

complex structure, but the RBD-CYN1 complex was unstable at high concentration,

which maybe contributed by the CYN1's high hydrophobic area or the 1: 2 binding

style to RBD. Nevertheless, we demonstrated that mimicking the small interface of

ACE2 has the potential to achieve broad-spectrum capability. We expected that this can

be improved with constantly improved AI-based protein design techniques so that the

interface can be more faithfully captured.

## Author Contribution

**Jian Zhan and Yaoqi Zhou:** Conceptualization, Investigation, Supervise, Writing-
review & editing. **Qinghui Nie:** Protein expression and purification, *E. coli* display to
evolve ACE2 mimic's affinity, Determined thermostability and affinity, Writing the
original draft, Writing- review & editing. **Hongmei Yin**: Sequence design of the
mimics, Writing the original draft. **Ke Chen**: Design pipeline development, backbone
design of the ACE2 mimics, Writing the original draft. **Chao Zhao**: Performed
computational simulations and binding analysis, Writing the original draft. **Weiwei**
**Li:** sample preparation for structure determination. **Jianxun Qi**: Supervising, Writing
review & editing**. Jinle Tang:** Supervise, Writing- review & editing. **John Z.H.**
**Zhang:** Supervise, review & editing.

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568
569 **Conflict of Interest**
570
571 All authors declare no financial interest. Zhan and Zhou are the CEO and the chair of
572 the scientific advisor board for Ribopeutic, respectively.
573

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