

Nanobody repertoire generated against the spike protein of ancestral SARS-CoV-2 remains efficacious against the rapidly evolving virus

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Abstract

To date, all major modes of monoclonal antibody therapy targeting SARS-CoV-2 have lost significant efficacy against the latest circulating variants. As SARS-CoV-2 omicron sublineages account for over 90% of COVID-19 infections, evasion of immune responses generated by vaccination or exposure to previous variants poses a significant challenge. A compelling new therapeutic strategy against SARS-CoV-2 is that of single domain antibodies, termed nanobodies, which address certain limitations of monoclonal antibodies. Here we demonstrate that our high-affinity nanobody repertoire, generated against wild-type SARS-CoV-2 spike protein (Mast, Fridy et al. 2021), remains effective against variants of concern, including omicron BA.4/BA.5; a subset is predicted to counter resistance in emerging XBB and BQ.1.1 sublineages. Furthermore, we reveal the synergistic potential of nanobody cocktails in neutralizing emerging variants. Our study highlights the power of nanobody technology as a versatile therapeutic and diagnostic tool to combat rapidly evolving infectious diseases such as SARS-CoV-2.

eLife assessment

This study presents **important** insights on the impact of SARS-CoV-2 variants on the binding and neutralization of a small library of nanobodies. The authors should be applauded for their comprehensive in vitro and in silico analyses of nanobody targeting of SARS-CoV-2 variants. The evidence supporting the claims of the authors is now **convincing**. This work will be of great interest to researchers in the fields of antibody/nanobody engineering and SARS-CoV-2 therapeutics.

Introduction

SARS-CoV-2 has infected >40% of the world's population (Collaborators 2022 [↗](#)) resulting in a devastating loss of life. As the SARS-CoV-2 pandemic enters its endemic phase (Meng, Irwin et al. 2023 [↗](#), Pilz and Ioannidis 2023 [↗](#)), multiple new variants continue to circulate. Since its initial spread, the rapid adaptation of the virus to selective pressures continues to produce variants of concern (VoC), of which the omicron variants presently account for over 90% of current SARS-CoV-2 infections (www.cdc.gov [↗](#)). SARS-CoV-2 displays three structural proteins that are potential targets for therapeutic intervention, but the primary focal point of vaccine development and many therapeutic strategies is the spike surface glycoprotein, which the virus uses to gain cell entry by attaching to the host cell angiotensin-converting enzyme 2 (ACE2) receptor (Jackson, Anderson et al. 2020 [↗](#), Krammer 2020 [↗](#), Letko, Marzi et al. 2020 [↗](#), Polack, Thomas et al. 2020 [↗](#)). The spike protein trimer consists of three domains: the receptor binding domain (RBD) on S1 that binds ACE2, the S1 N-terminal domain (NTD) that has a poorly defined function, and the S2 domain that is involved in virus-host cell membrane fusion (Walls, Park et al. 2020 [↗](#), Jackson, Farzan et al. 2022 [↗](#)). Glycosylation is most extensive on the NTD and the S2 domain, whereas the RBD is largely glycan free (Watanabe, Allen et al. 2020 [↗](#), Zhao, Praissman et al. 2020 [↗](#)). Consequently, it is unsurprising that the most antigenic domain on spike is the RBD, where the majority of neutralizing antibodies have been shown to bind. A comprehensive mapping of the epitopes from 1,640 neutralizing monoclonal antibodies (mAbs), all targeting the RBD, revealed 12 epitope groups (Cao, Yisimayi et al. 2022 [↗](#)). This data, combined with previous studies mapping the epitopes of antibodies targeting spike, reveal a total of 19 mAb epitope groups, including seven on the NTD (Wang, Muecksch et al. 2022 [↗](#)). Very few anti-S2 antibodies have been shown to be effective therapeutic options (Wec, Wrapp et al. 2020 [↗](#)), likely due to the shielding effect of S2 glycans (Grant, Montgomery et al. 2020 [↗](#)).

The mechanism by which omicron variants of SARS-CoV-2 (e.g., BA.1, BA.4, BA.5 and XBB) escape the neutralizing abilities of antibodies generated against spike proteins from preceding variants, whether by vaccination or infection, is largely attributed to the extensive number of mutations accumulated in spike (Greaney, Loes et al. 2021 [↗](#), Starr, Greaney et al. 2022, Dadonaite, Crawford et al. 2023 [↗](#)). Compared to wild-type SARS-CoV-2 spike, omicron BA.1 spike has 37 amino acid residue differences, with almost half located in the RBD domain (Mannar, Saville et al. 2022 [↗](#)). The omicron BA.4/BA.5 variants, which have identical spike proteins (Tegally, Moir et al. 2022 [↗](#)), have additional mutations (including the L452R substitution first seen in the delta variant), that render many previously broadly neutralizing antibodies ineffective (Cao, Yisimayi et al. 2022 [↗](#), Hachmann, Miller et al. 2022 [↗](#), Wang, Guo et al. 2022 [↗](#)). Of the monoclonal antibodies (mAbs) that previously received emergency use authorization (EUA) by the FDA for the treatment of SARS-CoV-2 infection, even cilgavimab and bebtelovimab that were respectively moderately and highly efficacious against omicron BA.5, are no longer effective against the current circulating variants XBB, BQ.1.1 and related sublineages (Takashita, Yamayoshi et al. 2022 [↗](#), Focosi, Quiroga et al. 2023 [↗](#), Imai, Ito et al. 2023 [↗](#)). As a result, no mAb therapy is currently approved by the FDA for treatment of SARS-CoV-2 infection (www.fda.gov [↗](#)).

Nanobodies, single-domain antibodies derived from a unique heavy chain-only class of llama antibodies, present numerous therapeutic benefits compared to mAbs. Their smaller size and increased stability make them more resistant to denaturation, simpler to produce, and easier to modify in order to adjust properties such as immunogenicity and half-life (Muyldermans 2013 [↗](#)). One potential advantage for neutralizing the spike protein is their compact size and distinctive binding attributes, which allow them to access and bind to epitopes that mAbs cannot reach. Consequently, while the antigenic evolution of the spike protein in response to antibodies has largely rendered mAbs ineffective in a therapeutic context, it remains uncertain how this applies to nanobodies. Moreover, the diminutive size of nanobodies enables them to bind concurrently to

a single antigen through non-overlapping epitopes, making them well-suited for creating nanobody mixtures with the potential for highly synergistic effects (Fridy, Li et al. 2014 [↗](#), Mast, Fridy et al. 2021 [↗](#)).

Here, we demonstrate that a subset of our previously published repertoire of nanobodies, generated against spike from the ancestral SARS-CoV-2 virus (Mast, Fridy et al. 2021 [↗](#)), retains binding and in vitro neutralization efficacy against circulating variants of concern (VoC), including omicron BA.4/BA.5. We show the power of nanobodies when working synergistically to create potent neutralizing mixtures against the different VoCs. We also predict that a subset of these nanobodies will remain efficacious against the circulating XBB and BQ.1.1 sublineages. Our study underscores the importance and versatility of large, diverse repertoires of nanobodies, in their potential to create long-term therapeutic options against rapidly evolving infectious agents such as the SARS-CoV-2 virus.

Results and Discussion

Nanobodies generated against wild-type SARS-CoV-2 spike remains efficacious against delta, and omicron lineages BA.1, BA.4/BA.5, XBB and BQ.1.1

From the original nanobody repertoire that we generated against SARS-CoV-2 wild-type spike protein (Mast, Fridy et al. 2021 [↗](#)), representative nanobodies from all 10 structurally mapped epitope groups that we previously identified (Mast, Fridy et al. 2021 [↗](#), Cross, Fridy et al. 2023 [↗](#)), were selected for SARS-CoV-2 pseudovirus (PSV) neutralization assays against the SARS-CoV-2 delta and omicron BA.1 strains (Figure 1 [↗](#), Table 1 [↗](#)). Of the 41 nanobodies tested, 35 remained efficacious against at least one variant, where 28 neutralized delta, 23 neutralized omicron BA.1, and 15 neutralized both. The RBD groups I, I/II, II, I/IV and IV – where nanobodies whose epitopes could not be distinguished between two groups are demarcated I/II and I/IV – and the anti-S2 groups (groups IX and X) contain a high number of nanobodies that neutralized delta. We note that IC₅₀s are not directly comparable across different experimental set-ups because measured values are highly dependent on the experimental conditions. For this reason, we included other published nanobodies as benchmarks in our original publication (Mast, Fridy et al. 2021 [↗](#)) and have subsequently maintained standard experimental conditions. Additionally, for computational epitope modeling, we selected nanobody candidates using a series of experimentally obtained structural restraints, as described in Mast, Fridy et al. 2021 [↗](#).

These observations align with data from mAbs approved by the FDA, where nanobody epitopes from groups I, II and IV (Figure 1B [↗](#)) overlap with the epitopes of five mAbs effective against delta, categorized into two of three classes on RBD: class 1 (etesivimab, casirivimab and amubarvimab) (Greaney, Starr et al. 2021 [↗](#), Li, Xue et al. 2021 [↗](#), Planas, Veyer et al. 2021 [↗](#), Takashita, Yamayoshi et al. 2022 [↗](#), Cox, Peacock et al. 2023 [↗](#)) and class 3 (imdemivab and bebtelovimab) (Cox, Peacock et al. 2023 [↗](#)) (Figure 2 [↗](#)). Notably, most of our group I and II nanobodies neutralized delta better than wild-type (Figure 1 [↗](#)), paralleling observations made with the mAb etesivimab (Wang, Li et al. 2022 [↗](#), Cox, Peacock et al. 2023 [↗](#)). In comparison, groups I, I/II, I/IV, V, VII, VIII and the anti-S2 nanobodies contained the majority of omicron BA.1 neutralizers, though here the neutralization potency of many nanobodies was generally decreased tenfold compared to wild-type. This decrease in neutralization potency largely correlates with the accumulation of omicron BA.1 specific mutations throughout the RBD, which potentially alter the nanobodies' binding sites, weakening their interaction with the BA.1 spike (Figure 1B [↗](#)). Concomitantly, groups I, I/II, I/IV and the anti-S2 groups contain nanobodies able to neutralize both delta and omicron BA.1. These results demonstrate the effectiveness of our original nanobody cohort against delta and omicron BA.1, targeting all major regions of spike.

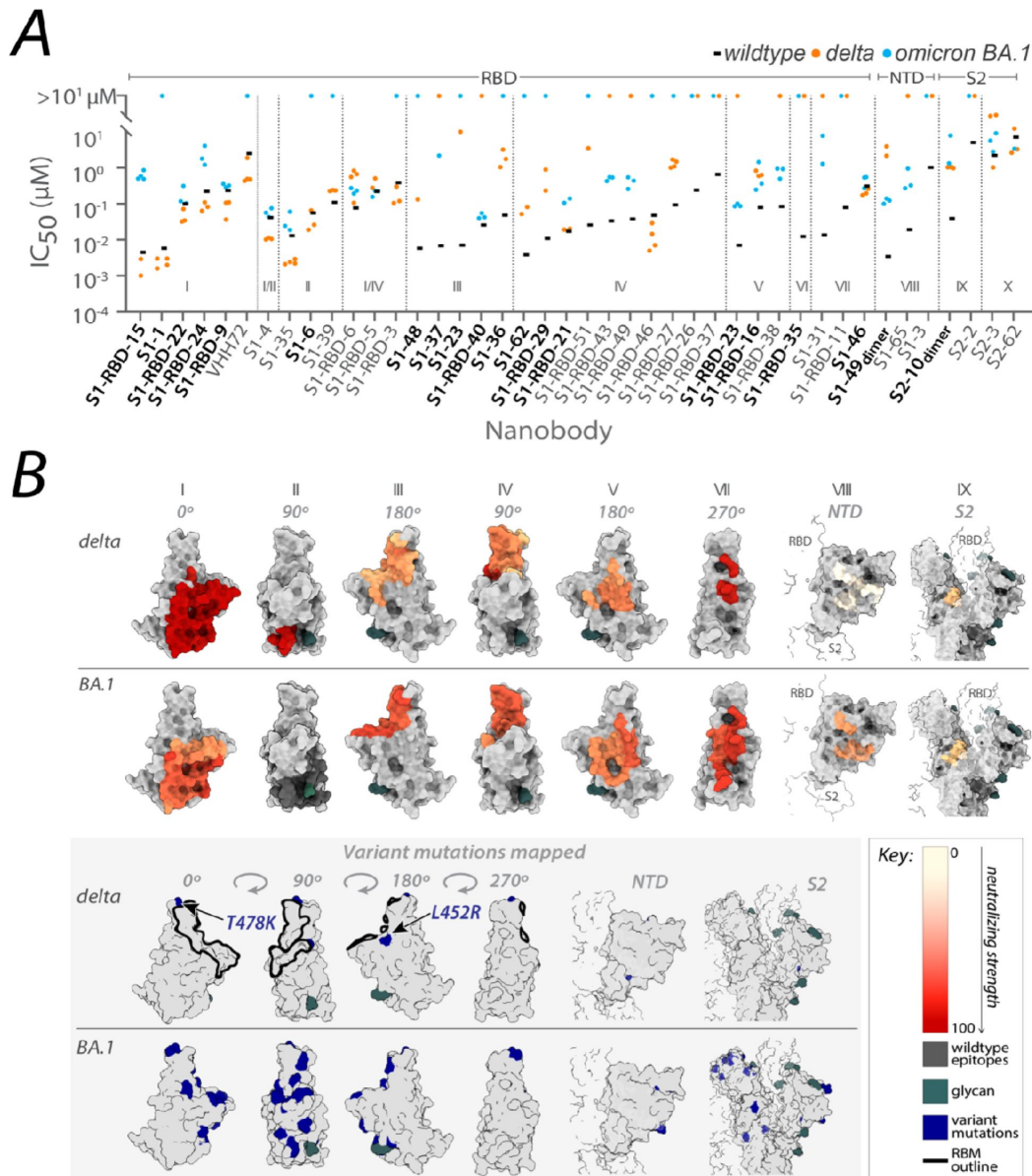


Figure 1.

Nanobody repertoires generated against wild-type SARS-CoV-2 remains efficacious.

Nanobodies targeting the S1-RBD, S1 non-RBD, and S2 regions of spike effectively neutralize lentivirus pseudotyped with delta and omicron BA.1 SARS-CoV-2 spikes (PSV) from infecting ACE2 expressing HEK293T cells. (A) The half-maximal inhibitory concentration (IC₅₀) is reported for the indicated nanobodies against wild-type (Mast, Fridy et al. 2021), delta, and omicron BA.1 PSV. These values are summarized in **Table 1**. Nanobodies are grouped by epitope and arranged within each epitope by neutralization efficacy against the wild-type PSV. $n \geq 4$ (B) The structural differences in the RBD of the delta (PDB ID: 7SBO) and omicron BA.1 (PDB ID: 7T9K) variants are depicted. Nanobody epitopes are heat-mapped ranging from pale white (epitopes with weak neutralization against SARS-CoV-2) to dark red (indicating strong neutralization). Boxed in grey are mutations specific to each variant mapped in blue on the aforementioned structures. The nanobodies that contributed to epitope mapping are in bold in panel A. The color bar scale for each epitope is the neutralizing strength of each nanobody epitope, calculated as the normalized $-\log_{10}$ ratio of nanobody binding (IC₅₀) to variant versus wild-type SARS-CoV-2 Spike S1. For groups with multiple nanobodies, the average $-\log_{10}$ (IC₅₀) is first calculated for the nanobodies within that group, then normalized to a neutralization score within the 0–100 range using the min and max average $-\log_{10}$ (IC₅₀) for that group. A higher score indicates more potent neutralization of the variant relative to the wild type. All structural representations were created on ChimeraX (Pettersen, Goddard et al. 2021).

Group #	Nanobody	wildtype ^d				Delta					Omicron BA.1					Omicron BA.4/5				Omicron XBB			Omicron BQ.1.1					
		K _{on}	K _{off}	K _D	PSV	K _{on}	K _{off}	K _D	PSV	IC ₅₀	change	K _{on}	K _{off}	K _D	PSV	IC ₅₀	change	K _{on}	K _{off}	K _D	IC ₅₀	K _{on}	K _{off}	K _D	K _{on}	K _{off}	K _D	
		(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)	IC ₅₀	(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)	IC ₅₀	Wildtype/Delta	(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)	IC ₅₀	Wildtype/Omicron	(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)	(nM)	(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)	(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)	(M ⁻¹ s ⁻¹)	(s ⁻¹)	(M)
1	S1-1	6.50E+05	5.98E-07	9.20E-12	6.7(1.0)	1.27E+06	1.00E-06	7.87E-13	2.3(0.5)	2.91	No binding			NA	No binding				Not tested			Not tested						
	S1-RB D-9	2.85E+05	1.23E-04	4.30E-10	235(97.5)	(1)3.93E+05	(1)1.00E-03	5.69E-10*	85.2	2.76	6.05E+03	1.82E-04	3.01E-08	323	0.73	4.36E+05	5.49E-03	1.26E-08	966	6.31E+05	5.41E-03	8.58E-09	2.65E+06	1.08E-02	4.06E-09			
					(2)1.76E-03	(2)5.04E-04		(24.3)						(20.8)					(220)									
	S1-RB D-15	7.52E+06	4.95E-04	6.58E-11	4.6(1.2)	2.55E+06	1.08E-04	4.25E-11	2(0.9)	2.30	1.95E+06	1.32E-02	6.79E-09	617(87)	0.01	No binding				Not tested			Not tested					
	S1-RB D-22	9.24E+05	4.42E-04	4.78E-10	100(0.1)	(1)1.07E+06	(1)6.25E-04	1.83E-10*	46.4(12)	2.16	7.30E+05	1.91E-04	2.62E-10	179(68)	0.56	1.07E+06	1.05E-02	9.80E-09	976	4.22E+05	3.61E-03	8.57E-09	5.66E+05	3.09E-03	5.46E-09			
					(2)9.10E-04	(2)4.15E-04													(285)									
	S1-RB D-24	2.12E+06	1.22E-03	5.75E-10	221(4)	(1)5.74E+05	(1)2.62E-03	1.64E-09*	85.5	2.58	(1)2.66E+05	(1)1.49E-02	1.34E-08*	238	0.02	9.25E+05	1.62E-02	1.75E-08		Not tested			Not tested					
					(2)1.61E-03	(2)9.00E-04		(13.5)			(2)2.63E-03	(2)8.24E-04		(878)														

Table 1.

Nanobody binding and neutralization characterization; related to Figure 2 [↗](#).

2	S1-6	5.9 2E+05	3.6 9E-04	6.2 2E-10	56. 1(2 0.7)	3.19 E+05	2.48 E-04	7.7 7E-10	37.3 (14. 7)	1.5 0	No binding				No binding			Not tested			Not tested				
	S1-35	2.7 0E+06	9.7 7E-05	3.6 2E-11	12. 5(0. 1)	5.83 E+05	1.48 E-05	2.5 4E-11	28(5 .2)	0.4 5	5.20 E+05	7.56 E-05	1.4 5E-10	49(13. 6)	0.2 5	1.02 E+06	4.35 E-05	4.2 7E-11	182 (55)	9.29 E+05	2.85 E-04	3.07 E-10	1.21 E+06	7.45 E-05	6.17 E-11
	S1-39	1.6 8E+06	1.0 6E-03	6.3 0E-10	111 (4.0)	2.30 E+06	6.73 E-04	2.9 2E-10	236(3.8)	0.4 7	(1)1. 14E +05	(1)5 .48 E-03	3.6 2E-09*	NA		7.59 E+04	7.30 E-04	9.6 1E-09		No binding			No binding		
3	S1-23	1.0 9E+06	1.0 7E-04	9.7 8E-11	5.7(2.2)	2.33 E+05	2.85 E-03	1.2 3E-08	NA		No binding			NA	No binding			No binding			No binding				
	S1-36	7.8 7E+06	1.7 2E-03	2.1 8E-10	48. 5(2 1.1)	(1)7. 90E +05	(1)2 E-02	9.0 0E-09*	1387 (209)	0.0 3	(1)1. 19E +06	(1)1 .64 E-02	3.4 7E-09*	NA	No binding			Not tested			Not tested				
	S1-RB-D-40	7.4 7E+05	2.7 7E-05	3.7 1E-11	25. 6(5. 9)	No binding			NA		2.25 E+05	3.71 E-05	1.6 5E-10	46. 1 (4.7)	0.5 5	(1)2. 54E +04	(1)3 .07 E-01	4.1 7E-06*	NA <10 uM	4.35 E+05	1.24 E-02	2.86 E-08	1.24 E+06	6.11 E-03	4.92 E-09
	S1-37	4.0 3E+06	2.7 5E-04	6.8 2E-11	6(8 0.7)	1.18 E+05	1.35 E-03	1.1 4E-08	NA		1.38 E+06	1.99 E-02	1.4 4E-08	NA	No binding			Not tested			Not tested				
	S1-48	2.6 1E+06	6.2 2E-05	2.3 9E-11	5.8 2(0. 5)	5.74 E+05	3.54 E-03	6.1 8E-09	145(5)	0.0 4	No binding			NA	No binding			Not tested			Not tested				
	S1-RB-D-21	3.1 5E+06	1.7 1E-03	5.4 5E-10	17. 3(3. 1)	(1)1. 82E +06	(1)1 .98 E-03	3.1 8E-10*	21(1 .3)	0.8 2	(1)1. 90E +05	(1)3 .16 E-03	5.8 0E-09*	113 (14)	0.1 5	No binding			Not tested			Not tested			
4	S1-RB-D-27	6.1 9E+06	1.2 4E-02	2.0 0E-09	163 (71. 4)	(1)4. 06E +05	(1)1 .48 E-02	8.2 4E-09*	1390 (198)	0.1 2	No binding			NA	No binding			Not tested			Not tested				
	S1-RB-D-29	5.3 6E+05	1.3 5E-03	2.5 1E-09	9.5 3(1. 0)	(1)1. 80E +05	(1)6 .99 E-03	1.6 9E-08*	564(332)	0.0 2	3.30 E+04	1.20 E-04	3.6 3E-09	NA	No binding			Not tested			Not tested				
	S1-RB-D-05	6.2 1E+05	1.8 2E-04	2.9 2E-10	33. 6(1. 3)	5.39 E+06	1.19 E-03	2.2 1E-10	NA		4.79 E+05	4.23 E-04	8.8 3E-10	103 2(5 43)	0.0 3	No binding			Not tested			Not tested			

Table 1. (continued)

43	S1-RB D-46	4.6 9E+05	5.7 9E-04	1.2 3E-09	48	1.11 E+06	3.12 E-04	2.8 1E-10	8.7(3)	5.5 2	No binding	NA	(1)2.17E+05	(1)1.38E-01	1.3 1E-07*	1020	2.84 E+05	5.06 E-02	1.78 E-07	1.90 E+05	3.65 E-02	1.92 E-07			
	S1-RB D-49	3.1 5E+05	5.3 4E-04	1.6 9E-09	37.6(5.6)	No binding					2.54 E+05	3.48 E-04	1.3 7E-09	404 (145)	0.0	No binding	Not tested			Not tested					
5	S1-RB D-16	1.6 8E+04	6.2 5E-05	3.7 3E-09	79.2(4.2)	3.13 E+04	3.04 E-04	9.7 4E-09	688 (72.6)	0.1 2	No binding		690 (150)	0.1	No binding		850 0(2300)			Not tested					
	S1-RB D-23	2.8 9E+06	4.6 1E-05	1.5 9E-11	7.3 1(0.4)	No binding			NA		7.91 E+05	5.43 E-05	6.8 7E-11	91.7 (5.2)	0.0	No binding	Not tested			Not tested					
	S1-RB D-38	1.1 2E+06	9.8 4E-04	8.7 9E-10	84.6(2.7)	No binding			NA		2.37 E+05	7.06 E-04	2.9 8E-09	790 (146)	0.1	No binding	NA <10 uM	Not tested			Not tested				
6	S1-RB D-35	1.3 3E+06	2.5 0E-04	1.8 8E-10	12.3(2.4)	No binding					No binding					No binding	Not tested			Not tested					
7	S1-46	2.2 2E+05	1.7 0E-04	7.6 6E-10	312 (14.0)	6.93 E+04	4.05 E-05	5.8 5E-10	210.2 (26)	1.4 9	2.12 E+05	1.24 E-04	5.8 5E-10	412 (116)	0.7 6	1.27 E+05	1.07 E-04	8.4 6E-10	NA <10 uM	5.01 E+04	1.88 E-04	3.75 E-09	5.44 E+04	1.99 E-04	3.65 E-09
8	S1-65	1.0 6E+06	1.6 7E-04	1.5 7E-10	7.3(6.0)	Not tested			NA		3.42 E+05	1.05 E-03	3.0 7E-09	517 (218)	0.0 1	No binding	Not tested			Not tested					
1/2	S1-4	1.2 6E+06	1.2 6E-04	1.3 7E-10	41.5(3.7)	1.38 E+06	4.83 E-05	3.5 1E-11	10.7 (0.4)	3.8 8	6.00 E+05	7.98 E-05	1.3 3E-10	58.6 (9.2)	0.7 1	1.32 E+06	1.32 E-03	1.0 0E-09	148.6 (37.6)	1.27 E+06	1.10 E-03	8.66 E-10	1.30 E+06	1.38 E-03	1.06 E-09
1/4	S1-RB D-3	5.1 2E+05	1.0 1E-02	1.9 8E-08	384 (18.7)	(1)3.27E+05	(1)5.54E-03	5.3 2E-09*	83.9	4.5 8	No binding					Not tested			Not tested						
	S1-RB D-5	7.2 1E+06	1.0 5E-03	1.4 5E-10	174 (3.3)	1.49 E+06	3.29 E-04	2.2 0E-10	280	0.6 2	(1)1.18E+05	(1)4.32E-03	7.7 5E-09*	204 (24.6)	0.8 5	1.19 E+06	9.71 E-03	8.1 9E-09		2.06 E+06	3.97 E-03	1.93 E-09	2.33 E+06	5.69 E-03	2.44 E-09
	S1-RB D-6	3.4 8E+06	1.1 3E-03	3.2 4E-10	77.2(2.1.8)	(1)6.68E+05	(1)5.04E-04	1.9 9E-10*	542.3 (155)	0.1 4	(1)1.43E+05	(1)3.44E-03	5.4 9E-9*	233.8 (23)	0.3 3	8.72 E+05	5.71 E-03	6.5 5E-09		2.25 E+05	5.52 E-03	2.46 E-08	2.25 E+05	4.94 E-03	2.19 E-08

#Values taken from (Mast, Fridy et al. 2021)

*Fit to a two-state mode

§S1-RBD-16 showed binding to omicron BA.1 and omicron BA.4/5 using ELISA

Table 1. (continued)

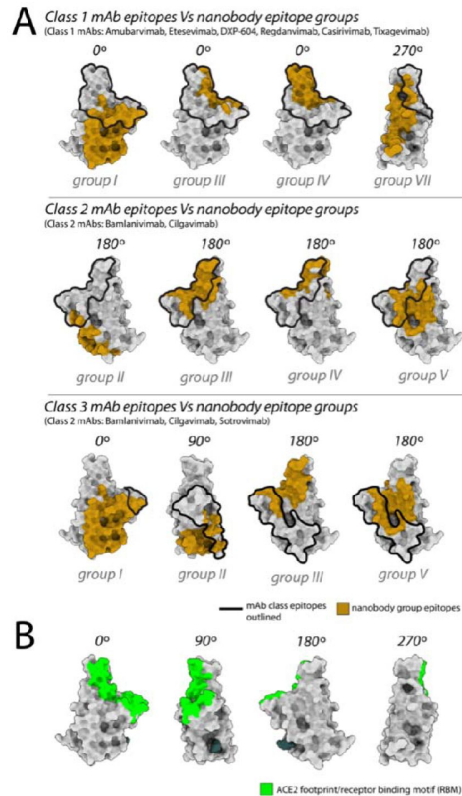


Figure 2.

Nanobody epitope groups and mAb epitope classes mapped on RBD.

(A) Nanobody epitope groups overlapping with the three mAb epitope classes (class 1, class 2 and class 3). Nanobody groups are highlighted in gold, while mAb class footprints are outlined in black. mAb epitopes are taken from Cox et al (Cox, Peacock et al.). (B) A single RBD subunit with the ACE2 footprint/RBM mapped in green. All epitopes are represented on the structure of wild-type RBD (PDB ID: 6M0J). All structure representations were generated using ChimeraX (Pettersen, Goddard et al.).

Of the nanobodies that neutralized both delta and omicron BA.1, representatives from each of the nanobody epitope groups were selected for SPR analysis, where S1 binders with mapped epitopes that neutralized one or both variants well were prioritized. SPR binding assessments to the spike S1 domain or RBD of delta revealed a pattern: nanobodies maintaining binding affinity generally also neutralized the virus with a statistically significant correlation between binding affinity and neutralization efficacy (Pearson's Correlation Coefficient: 0.71, p-value: 0.01; Spearman's rho: 0.63, p-value: 0.07). However, this correlation was not statistically significant for omicron BA.1 (Pearson's Correlation Coefficient: 0.27, p-value: 0.31) (**Figure 3A**, **Table 1**). Notably, while some nanobodies bound to the variants, they did not consistently neutralize them, suggesting additional factors influence neutralization beyond mere binding. We also tested this cohort for binding to omicron BA.4/BA.5 RBD using SPR, revealing that almost all the nanobodies in group I, I/II, II and I/IV retained binding to omicron BA.4/BA.5. Based on our previously mapped nanobody epitopes on spike (Mast, Fridy et al. 2021), these four groups appear to at least partially overlap with the spike epitope of the potent omicron BA.4/5 neutralizing mAb formerly FDA-approved, bebtelovimab (**Figure 2**) (Focosi, Quiroga et al. 2023). The nanobodies that retained binding to omicron BA.4/5 were further tested for binding against the omicron XBB and BQ.1.1 lineages in addition to S1-1 and S1-23, revealing nanobodies that bound omicron BA.4/BA.5, aside from S1-39, also bound omicron variants XBB and BQ.1.1 (**Figure 3A**, **Table 1**).

The remaining six nanobody groups tested, including the major nanobody groups III and IV, showed no detectable binding to omicron BA.4/BA.5/XBB/BQ.1.1 RBD, save for RBD-40 (group III) and RBD-46 (group IV), which exhibited a ~40-fold and ~20-fold decrease in affinity respectively, compared to wild-type. The diminished binding likely results from the concentration of omicron BA.4/BA.5/XBB/BQ.1.1 specific mutations that overlap with the epitope regions of the affected nanobodies. These mutations sufficiently alter the epitopes, to either abolish or significantly reduce binding (**Figure 3B**). Interestingly, one nanobody, S1-46 (group VII, **Figures 1B**; **3B**) retained wild-type binding affinity to RBDs from delta and all four omicron variants tested (**Figure 2**, **Table 1**). S1-46 binds a region on spike that is conserved across all variants to date, but which may be relatively inaccessible unless the RBDs are in the "up" conformation. The epitope of S1-46 is not targeted by any of the mAbs that previously received EUA by the FDA (Cox, Peacock et al. 2023). Guided by the SPR results against omicron BA.4/BA.5 RBD, nanobody neutralization of live BA.5 was performed using the plaque reduction neutralization test (PRNT) as previously described (Mast, Fridy et al. 2021). All five nanobodies tested neutralized the live omicron BA.5 live virus (**Figure 4**), corroborating our SPR observations. This indicates that our nanobody repertoire generated against wildtype spike has retained efficacy against omicron BA.4/BA.5.

Impact of spike structural differences across variants on nanobody binding and neutralization potency

The numerous structural differences observed in the spike protein of the delta and omicron sublineages compared to wild-type, has enabled these variants to escape the neutralizing effect of most mAbs, with only one clinically approved mAb retaining potency against omicron BA.4/BA.5 (Takashita, Yamayoshi et al., Cox, Peacock et al., Focosi, Quiroga et al.). These structural differences have also affected our nanobody cohort, likely playing a role in the differential binding and neutralizing abilities observed against the three tested variants (delta, omicron BA.1 and omicron BA.4/5). Notably, all but one of our nanobodies in groups I, I/II and II, displayed *better* neutralizing ability against delta compared to wild-type (**Figure 1**). It is possible that this difference is due to the delta spike trimer's preference for the up orientation of its RBDs, resulting from the increased dynamics in its S1 domain compared to wild-type (Wang, Liu et al. 2022); this more 'open' state likely allows greater accessibility to the epitopes of these nanobody groups, for net stronger binding. Additionally, the absence of delta-specific mutations within the epitope regions of groups I and II preserves their integrity for nanobody binding (**Figures 1B**; **3B**). For the remaining RBD nanobody groups, we are likely observing the impact of the two delta mutations T478K and

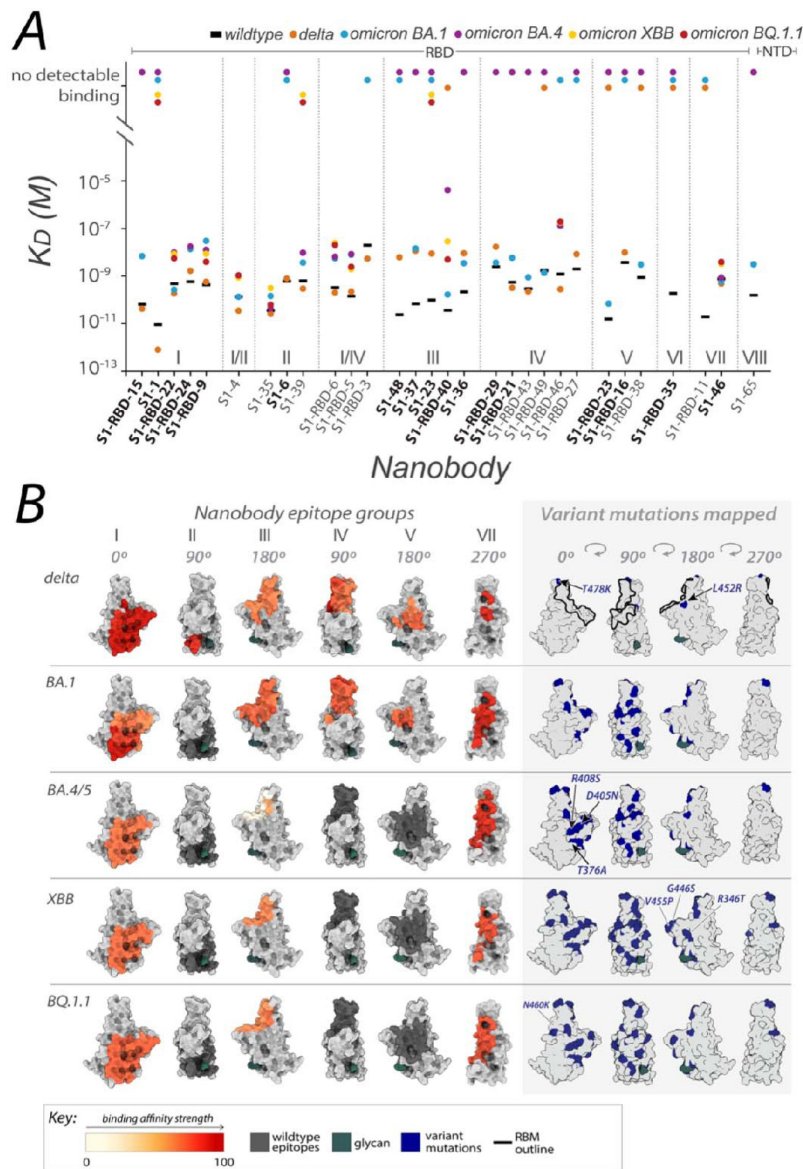


Figure 3.

Affinities of the nanobody repertoire against SARS-CoV-2 variants.

(A) Each nanobody is plotted against their affinity (K_D) for SARS-CoV-2 Spike S1 from wild-type, delta and omicron BA.1, BA.4, XBB and BQ.1.1 strains. Kinetic values are summarized in [Table 1](#). Nanobodies are characterized into their respective epitope groups as described previousl ([Mast, Fridy et al. 2021](#)). (B) Displayed are the structures of the RBD of spike delta (PDB ID: 7SBO), omicron BA.1 (PDB ID: 7T9K), omicron BA.4 RBD (modeled using AlphaFold), omicron XBB (PDB ID: 8IOU) and omicron BQ.1.1 (PDB ID: 8FXC). The structures feature heat-mapped epitopes of binding, ranging from pale white (weak binding to SARS-CoV-2) to dark red (strong binding to SARS-CoV-2). In the grey box, mutations specific to each variant are highlighted in blue. The nanobodies that contributed to epitope mapping are in bold in panel A. The color bar scale indicates each epitope's binding affinity strength, represented as the normalized $-\log_{10}$ ratio of nanobody binding (K_D) of variant versus wild-type SARS-CoV-2 Spike S1. For groups with multiple nanobodies, the average $-\log_{10}$ (K_D) for the nanobodies within that group were calculated, then normalized to an affinity score ranging from 0–100 using the min and max average $-\log_{10}$ (K_D) for that group. Higher $-\log_{10}$ ratios indicates stronger binding of the nanobody to the variant versus wild type. S1-RBD-16 bound omicron BA.1 and BA.4/5 in ELISA. S1-RBD-11 was not tested against omicron BA.4. S1-65 was not tested against BA.1. Only S1-1, S1-RBD-22, S1-RBD-9, S1-4, S1-35, S1-39, S1-RBD-6, S1-RBD-5, S1-23, S1-RBD-40, S1-RBD-46 and S1-46 were tested against omicron XBB and omicron BQ.1.1. All structure representations were generated using ChimeraX ([Pettersen, Goddard et al. 2021](#)).

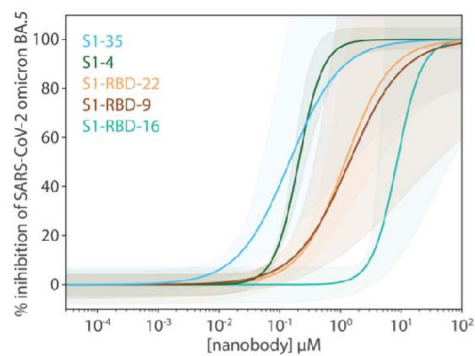


Figure 4.

Potent neutralization by broadly active nanobodies.

Nanobodies targeting the S1-RBD of spike and raised against the original wild-type sequence remain highly efficacious in neutralizing an evolved variant of SARS-CoV-2, omicron variant BA.5. The derived neutralization curves are plotted from the results of a plaque-forming reduction neutralization test with the indicated nanobodies. Serial dilutions of each nanobody were incubated with ~200 SARS-CoV-2 virions for 60 min and then overlaid on a monolayer of TMPRSS2-expressing Vero E6 cells. After 72 h, cells were fixed and stained with crystal violet stain (1% w/vol in 20% ethanol) allowing for the enumeration of viral plaques. The percent plaque inhibition for each nanobody dilution, summarized in Figure 4–Source data 1, was used to fit the neutralization curves depicted in the figure. The colored shaded areas denote 90% confidence intervals for each fitted curve. $n \geq 3$.

L452R on nanobody binding and neutralization. These two mutations lie within the mapped epitope regions of group III, IV and V nanobodies (**Figures 1B** [↗](#); **3B** [↗](#)), which may alter the epitopes of many of these nanobodies enough to negatively impact both binding and neutralization. Importantly, the L452R mutation seems to be a key substitution that contributes to reduced or abolished neutralizing abilities of many mAbs (Laurini, Marson et al. 2021 [↗](#), Starr, Greaney et al. 2021 [↗](#), Starr, Greaney et al. 2021 [↗](#)). Antibodies that rely on L452 to create hydrophobic interactions within their epitope will most likely have their binding greatly disrupted with a substituted arginine. Coupled with the stronger affinity between ACE2 and spike caused by L452R (Motozono, Toyoda et al. 2021 [↗](#), Yan, Hou et al. 2022 [↗](#)), this substitution can greatly lessen the neutralization ability of nanobodies and antibodies targeting this region.

Unlike wild-type, the spike trimer of omicron BA.1 favors a one-RBD up confirmation (Zhao, Zhou et al. 2022 [↗](#)). Though this conformation may facilitate access of our nanobodies to their epitopes on RBD, unlike delta, omicron BA.1 contains many unique mutations distributed throughout the RBD domain that overlap or flank the epitopes of almost all our nanobody groups (**Figures 1B** [↗](#); **3B** [↗](#)). These mutations likely alter the epitopes of our nanobodies and may be the major contributing factor to the observed decrease in binding affinity and neutralizing potency of our nanobodies against omicron BA.1 (**Figures 1A** [↗](#); **3A** [↗](#)). However, the following cohort of nanobodies retained binding and neutralizing ability similar to wild-type against omicron BA.1: S1-RBD-22, S1-RBD-9 (both Group I, Class 1), S1-4 (Group I/II, Class 3), S1-RBD-5 (Group I/IV, Class 3), S1-46 (Group VII) and members of Group X. The additional mutations on the spike of omicron BA.4/BA.5, many of which overlap with our nanobody epitopes, are predicted to further impact the binding of our nanobody repertoire to this variant. For example, omicron BA.4/BA.5 re-introduces the key L452R substitution from delta, which when combined with the accumulated mutations on spike from omicron precursor sub-lineages, could be responsible for the observed loss of binding of numerous nanobodies in groups III, IV and V. Furthermore, the differential binding and neutralizing abilities of group I and group II nanobodies against omicron variants BA.4/BA.5/XBB/BQ.1.1 may be because of the T376A, D405N and R408S substitutions, which lie within and near the epitope regions of groups I and II respectively (**Figure 3B** [↗](#)). Importantly, at least 5 nanobodies from groups I, I/II, II and V retained neutralization activity against omicron BA.5 (**Figure 4** [↗](#)), demonstrating the broad specificity of this set of nanobodies.

Nanobodies that lose neutralization ability can still bind spike

Our SPR experiments largely correlated with the neutralization data: nanobodies that showed binding to spike also neutralized the virus, and where the binding affinity decreased significantly, a loss of neutralization ability was observed. This was seen with the group III nanobodies S1-23 and S1-37 that both demonstrated significantly reduced binding affinity and neutralization potency against both delta and omicron BA.1. This reduction in efficacy is likely attributable to the L452R mutation against delta, and the extensive amino acid changes in omicron BA.1 compared to wild-type (see above). However, we observed instances where nanobodies retained binding to spike yet no longer neutralized the virus. Nanobodies S1-36, S1-39 and S1-RBD-29 showed binding to omicron BA.1 (**Figure 3A** [↗](#)), yet none neutralized the variant in the pseudovirus assay (**Figure 1A** [↗](#)). These nanobodies are in groups III, II and IV respectively (**Figure 3B** [↗](#)) – epitope regions that contain numerous omicron BA.1 mutations. For all three nanobodies, the mutations peppered throughout their epitope space have likely altered the binding landscape, resulting in decreased affinity (~100-fold for S1-36 and S1-39) and ineffective neutralization *in vitro*. Additionally, S1-RBD-29 binds an epitope that overlaps with the ACE2 binding site, and likely neutralizes the wild-type strain by blocking the ACE2 interaction (**Figure 1B** [↗](#)). Alterations in the omicron BA.1 epitope may have changed the orientation of the nanobody as it binds, negating effective blocking of ACE2 binding and thus neutralization. In contrast, S1-37 binds both delta and omicron BA.1 with similar reduced affinities (>100-fold decrease compared to wild-type), yet only neutralized omicron BA.1 (**Figure 1A** [↗](#)). As mentioned above, the epitope of S1-37 overlaps with the delta L452R mutation (**Figure 1B** [↗](#)), which has impacted the effectiveness of numerous neutralizing antibodies (Bian, Gao et al. 2021 [↗](#)). It is possible that this mutation alone drastically weakens S1-

37 binding to delta and consequently virus neutralization. Lastly, S1-RBD-43, whose epitope as a group IV binder is predicted to overlap with the ACE2 binding footprint (**Figures 1B** [↗](#); **3B** [↗](#)), showed binding to delta with equal affinity to wild-type (**Figure 3A** [↗](#)), yet does not neutralize delta (**Figure 1A** [↗](#)). The delta mutation T478K is present within the epitope region of group IV (**Figures 1B** [↗](#); **3B** [↗](#)), and was shown to significantly increase the interaction of delta spike for ACE2 by creating a new salt-bridge at the RBD / ACE2 interface (Cheng, Krieger et al. 2022 [↗](#)). It may be that, as suggested for S1-RBD-29, the binding orientation of S1-RBD-43 has been altered allowing the RBD to maintain an interaction with ACE2 despite the presence of S1-RBD-43, thus rendering the nanobody ineffective in neutralizing delta.

Identification of variant-specific epitopes and broadly neutralizing epitope groups

The results of our neutralization assays and affinity measurements revealed that ~1/3 of our original repertoire of 116 nanobodies (Mast, Fridy et al.) generated against wild-type SARS-CoV-2 remain effective binders/neutralizers of the variants tested. Specifically, nanobodies from 11 of the 18 nanobody groups (inclusive of the 10 mapped epitopes) demonstrated efficacy against one or more of the delta, omicron BA.1, omicron BA.4/BA.5, omicron XBB and omicron BQ.1.1 lineages. The varied efficacy of nanobodies within each group, along with structural modeling, enabled us to expand and further refine our original six structurally modeled RBD epitope groups to a total of 12 (**Figure 5A** [↗](#)). These 12 groups were then contrasted with the three mAb classes containing mAbs previously approved for EUA by the FDA (**Figure 2** [↗](#)). This comparison revealed the following overlaps between our 12 RBD nanobody epitope groups and the three mAb classes: groups I, III, IV and VII overlap with class 1 binders; group II, III, IV and V overlap with class 2 binders; and group II, III and V overlap with class 3 binders (**Figure 2** [↗](#)).

Interestingly, the epitope footprint of many of our nanobody classes extends beyond those of the three mAb classes (**Figure 2** [↗](#)). For example, the majority of the epitope regions of the group I, II, V and VII nanobodies do not overlap with the mAb classes and are not a binding/neutralizing hotspot for mAbs (Almagro, Mellado-Sanchez et al. 2022 [↗](#), Cao, Yisimayi et al. 2022 [↗](#)); instead, these epitopes extend away from the ACE2 binding site (**Figures 1B** [↗](#); **3B** [↗](#); **2** [↗](#)), as seen in particular with groups I, II and V nanobodies. These regions may be inaccessible to mAbs, possibly due to steric limitations, a property nanobodies readily overcome due to their small size. Additionally, our data allowed us to identify variant-specific epitope groups (**Figure 5B** [↗](#)), where we define variant-specific nanobodies as nanobodies that bind a single additional variant alongside the original Wuhan strain, summarized as follows: of the 26 nanobodies that showed binding/neutralization to delta, 5 were specific only for delta; of the 21 nanobodies that showed binding/neutralization to omicron BA.1, two were specific for omicron BA.1. Strikingly, we have in our cohort 8 nanobodies able to bind delta and the omicron lineages BA.1/BA.4/BA.5/XBB/BQ.1.1 (**Figure 5B** [↗](#)). We further predict these 8 nanobodies will be effective binders against current circulating strains of the virus including omicron EG.5 and HV.1 as the epitope regions (or predicted epitopes) of these nanobodies do not vary significantly from omicron lineages XBB and BQ.1.1 (**Figures 5C** [↗](#); **Figure 5-Figure supplement 1** [↗](#)).

Nanobody synergy involving a non-neutralizing nanobody

We previously established that cocktails of nanobodies exhibit enhanced resistance to mutational escape (Mast, Fridy et al. 2021 [↗](#)). Excitingly, not only was the barrier to mutational escape extremely enhanced, but for certain combinations of nanobodies, their mechanisms of neutralization were synergistic, providing far more potent neutralization in combination than expected from the neutralization by either nanobody alone (Mast, Fridy et al. 2021 [↗](#)). Our present observations of nanobodies that retained binding to variants of the spike RBD (**Table 1** [↗](#)) despite losing neutralization efficacy (**Figure 1A** [↗](#); **Table 1** [↗](#)) afforded us an opportunity to test whether the synergy observed for certain nanobody combinations was dependent on their ability to neutralize.

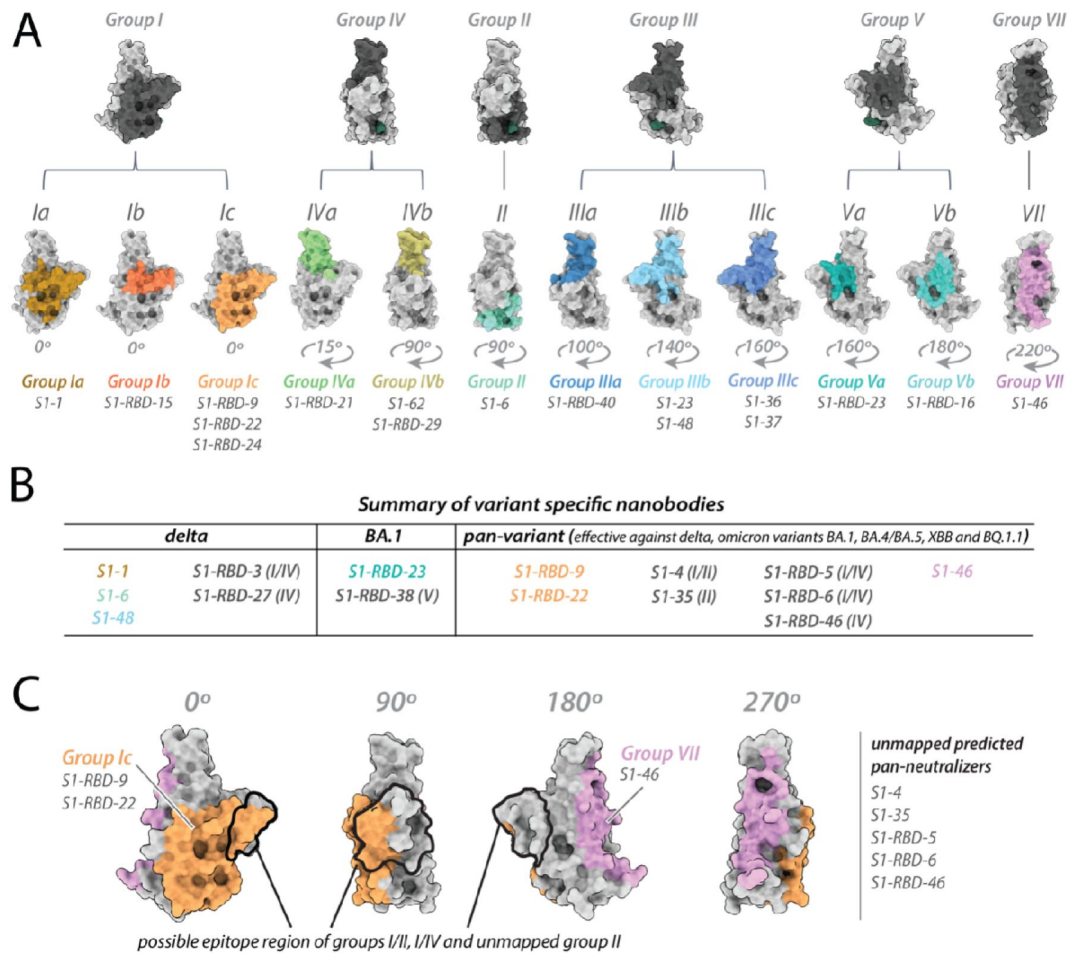


Figure 5.

Refining epitopes of the nanobody repertoire.

All epitopes are mapped on the structure of wild-type RBD (PDB ID: 6M0J). (A) The original six RBD nanobody epitope groups capable of binding and/or neutralizing one or more variants are highlighted in dark gray. Further refinement of four groups: I, III, IV and V led to the identification of six additional epitope groups – resulting in a total of 12 epitope groups able to bind one or more variants of concern. (B) Summary of variant specific and broadly reactive nanobodies. (C) Nanobody groups predicted to bind/neutralize the circulating omicron variants EG.5 and HV.1. All structure representations were created on ChimeraX (Pettersen, Goddard et al. 2021 <https://doi.org/10.26434/chemrxiv-2021-03>).

The synergistic S1-1 and S1-23 pair effectively neutralized the wild-type PSV with their epitopes on opposing surfaces of the RBD, permitting simultaneous binding and enhanced neutralization when delivered as a cocktail (Mast, Fridy et al. 2021 [↗](#)) (**Figures 1A [↗](#); 3A [↗](#); 6A [↗](#) panel i**). While S1-1 remained efficacious against the delta variant of SARS-CoV-2, the L452R mutation in the delta RBD likely negatively impacted S1-23 (discussed above), weakening its binding affinity by ~1000-fold (12 nM) and negating its neutralization efficacy at concentrations <10 μ M (**Figures 1A [↗](#); 3A [↗](#); 6A [↗](#) panel ii**). Surprisingly, when provided in combination with S1-1, which displays increased binding affinity and enhanced neutralization against the delta variant, S1-23 was able to further enhance the neutralization capabilities of S1-1, synergistically, at concentrations above 10^{-3} μ M (**Figure 6A [↗](#) panel ii**), by up to 42-fold. This synergistic interaction, however, did not apply to situations where extensive mutations are present in the RBD, such as in the omicron sublineages of SARS-CoV-2, which ablated the binding and neutralization efficacy of both S1-1 and S1-23 (**Figures 1A [↗](#); 3A [↗](#); 6A [↗](#) panel iii**).

We also tested the broadly neutralizing nanobody S1-RBD-22 in combination with S1-36 (**Figure 6B [↗](#)**). Like S1-23, the epitope of S1-36 is opposite that of S1-RBD-22, permitting simultaneous binding to a single RBD (Mast, Fridy et al. 2021 [↗](#)). However, while its neutralization efficacy dropped off when delivered to either delta or omicron BA.1 PSV alone, its ability to bind to its epitope was only marginally impacted (**Figures 1A [↗](#); 3A [↗](#); 6B [↗](#)**). When provided in combination with S1-RBD-22, S1-36 synergistically enhanced by up to 80-fold the neutralization efficacy of S1-RBD-22 against both delta and omicron BA.1 PSV (**Figure 6B [↗](#) panels ii, iii**). The synergies observed between S1-1 and S1-23, and between S1-RBD-22 and S1-36, appear to be specific rather than due to pleiotropic effects. This is evidenced by the lack of enhanced neutralization when S1-1 is combined with the nonspecific LaM2 nanobody against delta PSV (**Figure 6C [↗](#)**). Furthermore, in the case of S1-23, binding to its non-neutralizing epitope on the RBD of delta PSV was able to induce dose-dependent antagonistic effects on the neutralizing efficacy of S1-RBD-16, which binds to a neighboring epitope that can be competitively blocked by S1-23 (**Figure 6D [↗](#)**) (Mast, Fridy et al. 2021 [↗](#)).

Conclusions

Collectively, our binding and neutralization data allowed us to identify the regions of spike in multiple VoC that remained vulnerable to our original repertoire of nanobodies raised against wild-type SARS-CoV-2 spike (**Figure 7 [↗](#)**) (Mast, Fridy et al. 2021 [↗](#)). Unlike the epitope groups defining mAb binding sites on spike, many of our nanobody epitope groups remained efficacious in the neutralization of different VoC (**Figures 6 [↗](#); 7 [↗](#)**), possibly due to each nanobody's smaller epitope footprint allowing their access to regions of spike inaccessible to mAbs. However, the substantial changes on the surface of spike that has occurred as SARS-CoV-2 has evolved from one variant to the next has also negatively impacted many nanobodies by abolishing or weakening their binding and/or neutralization activity. This weakening is most evident for nanobodies directed against the NTD, and against the receptor binding motif that engages ACE2 (**Figures 1B [↗](#); 2B [↗](#)**).

Intriguingly, we discovered nanobodies that maintained binding capabilities while losing their neutralizing properties. This finding paves the way for engineering these nanobodies through approaches like oligomerization, which has proven effective in boosting neutralization (Wrapp, De Vlieger et al. 2020 [↗](#), Mast, Fridy et al. 2021 [↗](#)). Remarkably, we also showed that such non-neutralizing binders can nevertheless retain effectiveness as components of synergistic nanobody cocktails, suggesting their potential for widespread antiviral applications. This finding indicates that the mechanism of synergy can operate through epistatic interactions from binding alone, not solely through direct neutralization. Furthermore, these synergistic pairings hold promise as

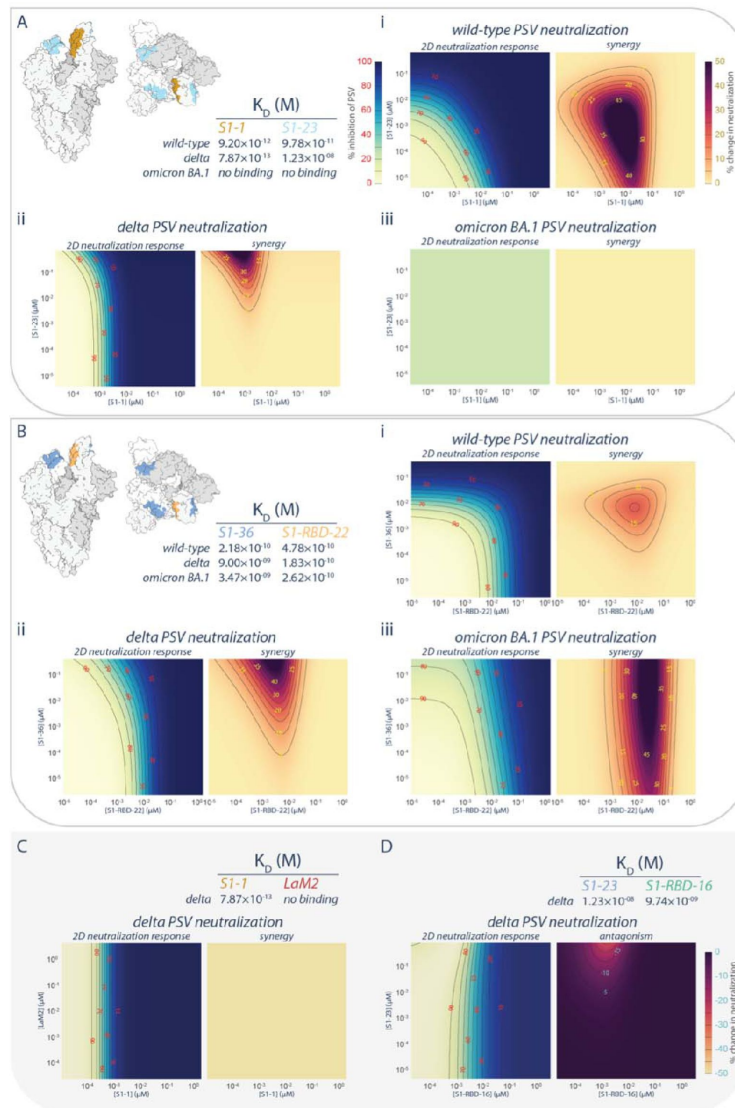


Figure 6.

Persistence of synergistic neutralization with nanobody cocktails against SARS-CoV-2 Variants of Concern.

(A) S1-1 synergizes with S1-23 in neutralizing SARS-CoV-2 PSV. The upper left panel shows two representations of spike with the accessible S1-1 (dark goldenrod) and S1-23 (sky blue) epitopes (PDB ID: 6VYB). The measured affinities for S1-1 and S1-23 (Table 1) for the RBD of wild-type, delta, and omicron BA.1 are displayed. Both S1-1 and S1-23 neutralize wild-type (i), whereas only S1-1 neutralizes delta at the concentrations shown (ii). In spite of a lack of neutralization at these concentrations, S1-23, synergizes with S1-1 and enhances its neutralization of delta SARS-CoV-2 PSV (ii). As neither S1-1 or S1-23 are able to bind to the RBD of omicron BA.1, neither nanobody neutralizes omicron BA.1 SARS-CoV-2 PSV (iii). (B) S1-36 synergizes with S1-RBD-22 in neutralizing SARS-CoV-2 PSV. As in A, the upper left panel shows two representations of spike with the accessible S1-36 (cornflower blue) and S1-RBD-22 (sandy brown) epitopes. The measured affinities for S1-36 and S1-RBD-22 are displayed. Both S1-36 and S1-RBD-22 neutralize wild-type (i), whereas only S1-RBD-22 effectively neutralizes delta and omicron BA.1 SARS-CoV-2 PSV at the concentrations shown (ii and iii, respectively). However, S1-36 synergizes with S1-RBD-22 and enhances its neutralization of the three depicted SARS-CoV-2 pseudoviruses (i, ii, and iii). (C) An example of no interactions (synergistic or antagonistic) between S1-1 and LaM2 (Fridy, Li et al. 2014), a non-specific nanobody that does not bind the RBD of delta. (D) An example of antagonism, where higher concentrations of S1-23 interferes with the ability of S1-RBD-16 to neutralize delta SARS-CoV-2 PSV. These nanobodies have adjacent epitopes on the RBD of spike and were previously shown to interfere with each other's binding to their respective epitope (Mast, Fridy et al. 2021). n = 4. Source data in Figure 6–Source data 2.

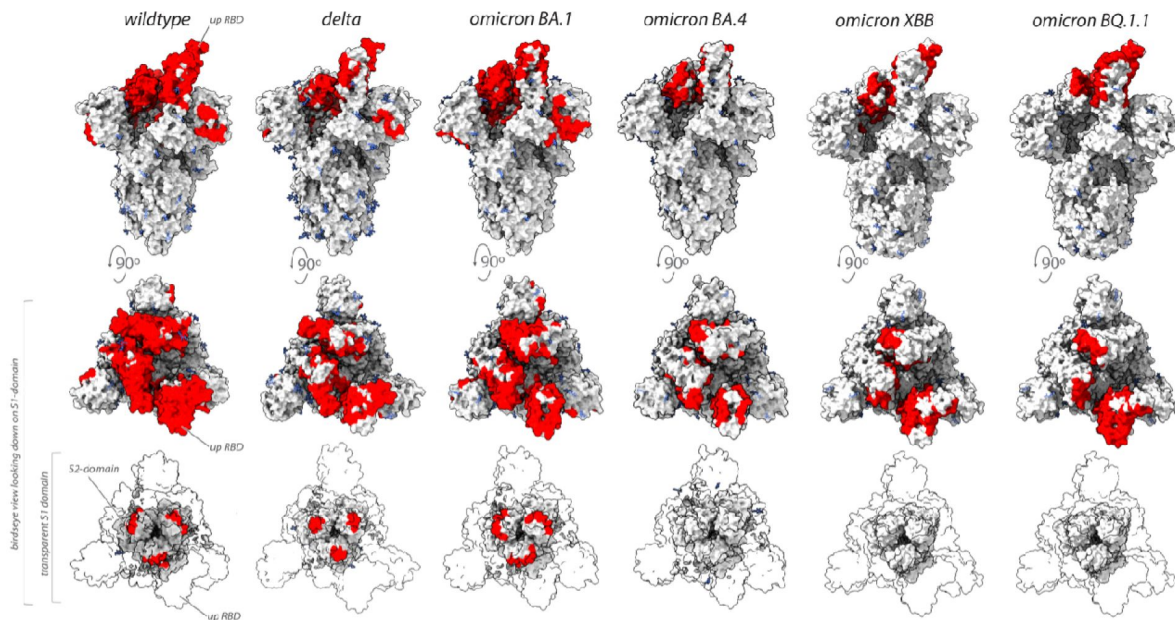


Figure 7.

Nanobodies effective against circulation variants of concern.

The nanobody epitopes (red) that retain effectiveness against wild-type (PDB ID: 7KNB), delta (PDB ID: 7V70), omicron BA.1, omicron BA.4 (the nanobody epitopes were mapped on to PDB ID: 7XO5 for both BA.1 and BA.4 due to the lack of a suitable BA.4 structure), omicron XBB.1 and BQ.1.1 (the nanobody epitopes were mapped on to PDB ID: 8IOU for both XBB.1 and BQ.1.1 due to the lack of a suitable BQ.1.1 structure). The spike trimer (*silver*) where glycans are represented as *light blue* sticks for each variant is displayed in three views: top = side-view of spike trimer; middle = birds-eye view looking down on the S1 domain and; bottom = birds-eye view (same as middle view), with the S1 domain rendered transparent to enable visualization of the S2-domain. All structure representations were created on ChimeraX (Pettersen, Goddard et al. 2021 [\[3\]](#)).

therapeutic options when formulated as multivalent hetero-oligomers and we propose that any of the nanobodies we have demonstrated to show pan-VoC activity would be prime candidates for further optimization.

In this study, we identified nanobodies that specifically recognize only certain SARS-CoV-2 variants (**Figure 5B** [↗](#)), allowing for the possibility of distinguishing between different VoC. By utilizing these nanobodies as molecular probes in diagnostic tests, a unique “molecular fingerprint” could define each variant based on the combinations of nanobodies that bind to and recognize the specific virus particle. Consequently, this approach could enable the accurate and rapid diagnosis of SARS-CoV-2 infections as well as provide real-time identification of the specific variant causing the infection, thus enhancing surveillance and tailoring treatment strategies accordingly to optimize patient outcomes and control the spread of the virus.

Together, our findings highlight the strength and variety of the heavy chain-only category of llama antibody immune responses and, as a result, the extensive repertoires of high-affinity nanobodies. This result emphasizes the distinct benefits of nanobody technology, which enables extensive coverage of antigenic regions while simultaneously targeting multiple unique epitope sites. This approach paves the way for investigating a wide range of therapeutic options against rapidly evolving proteins like the SARS-CoV-2 spike protein, ultimately aiding in our preparedness and defense against future pandemics or major outbreaks.

Materials and Methods

K_D measurements

Series S CM5 sensor chips (Cytiva) were immobilized with recombinant delta Spike S1, delta RBD and omicron RBD at 12.5 ug/mL, 10 ug/mL and 15 ug/mL respectively using EDC/NHS coupling chemistry according to the manufacturer’s instructions. K_D measurements were performed on a Biacore 8k (Cytiva) as previously described ([Mast, Fridy et al. 2021](#) [↗](#)), using 5-8 concentrations of each nanobody. Data was processed and analyzed using the Biacore Insight Evaluation software.

Structural modeling

Integrative structure modeling of nanobody epitopes on the different variant spike proteins, proceeded through the standard four-stage protocol ([Kim, Fernandez-Martinez et al. 2018](#) [↗](#), [Webb, Viswanath et al. 2018](#) [↗](#), [Rout and Sali 2019](#) [↗](#), [Sali 2021](#) [↗](#), [Saltzberg, Viswanath et al. 2021](#) [↗](#)). This protocol was implemented using the *Python Modeling Interface* package, a library for modeling macromolecular complexes based on the open-source *Integrative Modeling Platform* software, version 2.15.0 (<https://integrativemodeling.org> [↗](#)). Only (a subset of) nanobodies with pre-determined experimental escape mutations on the Wuhan spike structure were selected for modeling. Separate models were computed for rigid-receptor-rigid ligand-type binary docking of representative nanobodies from Group-1 (S1-1, S1-RBD-[9, 15, 22, 24]), Group-2 (S1-6), Group-3 (S1-[23, 36]), Group-4 (S1-RBD-[21, 29]), Group-5 (S1-RBD-[16, 23]), Group-7 (S1-46), Group-8 (S1-49) and Group-9 (S2-10), to the variant spike structures. S1-49 was docked to a monomeric S1-NTD domain, S2-10 was docked to the trimeric S2 (ecto-) domain, while the remaining nanobodies were docked to the S1-RBD domain. Monomeric S1-RBD spanning amino acid residues 333-526 was represented using the 2.45 Å crystal structure of the ACE2 bound RBD (6M0J.E; ([Lan, Ge et al. 2020](#) [↗](#))) for the original virus, the 4.30 Å cryo-EM structure of up-RBD pre-fusion spike (7SBO.A; ([Zhang, Xiao et al. 2021](#) [↗](#))) for the delta variant, the ACE2 bound 2.45 Å cryo-EM spike structure (7T9K.B; ([Mannar, Saville et al. 2022](#) [↗](#))) for the omicron BA.1 variant and a structure predicted using AlphaFold-2 for the omicron BA.4/BA.5 variant, the 3.18 Å cryo-EM structure of ACE2 bound up-RBD pre-fusion spike (8IOU.A; ([Tamura, Ito et al. 2023](#) [↗](#))) for the XBB.1 variant, and the 3.20 Å cryo-EM structure

of the BQ.1.1 variant RBD (8FXC:E; (Tamura, Ito et al. 2023 [↗](#))) in complex with ACE2 and the S309-neutralizing antibody Fab fragment. Missing residues in 8FXC:E were filled in using Modeller (Webb and Sali 2016 [↗](#)).

For the original virus, monomeric S1-NTD, spanning amino acid residues 16-305, was represented using the crystal structure of the S2M28 Fab bound NTD (7LY3.A; (McCallum, De Marco et al. 2021 [↗](#))), while trimeric S2 was represented using the amino residues 689-1162 (for each monomer) from the 2.9 Å cryo-EM structure 6XR8 (Cai, Zhang et al. 2020 [↗](#)). NTD and S2 structures for delta and omicron variants were extracted from the corresponding whole spike structures.

Structural models for all 15 nanobodies and the omicron BA.4 RBD were built with the ColabFold implementation of AlphaFold2 (Jumper, Evans et al. 2021 [↗](#), Mirdita, Schutze et al. 2022 [↗](#)). The protocol included automatic refinement of the CDR loops through an all-atom energy minimization of the AlphaFold2-predicted structure using the AMBER molecular mechanics force field (Hornak, Abel et al. 2006 [↗](#)). We verified that the predicted nanobody structures are within 3-4 Å backbone RMSD from the comparative models of these nanobody sequences published previously (Mast, Fridy et al. 2021 [↗](#)). The CDR region boundaries in the nanobody structures were assigned using the FREAD algorithm as implemented within the SabPred web server (Rausch 1991 [↗](#)).

To make structural sampling sufficiently efficient, the system was represented at a resolution of one bead per residue, and the receptors and all nanobodies were treated as rigid bodies. For each nanobody, alternate binding modes were scored using spatial restraints enforcing receptor-ligand shape complementarity, cross-link satisfaction and proximity of CDR3 loops on the nanobodies to escape mutant residues on the corresponding receptor. With the receptor fixed in space, 1,200,000 alternative docked nanobody models were produced through 20 independent runs of replica exchange Gibbs sampling based on the Metropolis Monte Carlo algorithm, where each Monte Carlo step consisted of a series of random rotations and translations of rigid nanobodies. The initial set of models was filtered to obtain a random subsample of 30,000 models, which were clustered by the structural similarity of their interfaces to the receptor; this similarity was quantified by the fraction of common contacts (fcc) between receptor and nanobody was used to characterize interface similarity between alternate nanobody poses (Rodrigues, Trellet et al. 2012 [↗](#)). Binding poses from the most populated cluster were selected for further analysis. Five independent random subsamples of 30,000 models each were generated from the set of all models post-structural sampling, and the entire protocol of interface similarity-based clustering and top cluster selection was repeated each time. No significant differences among these five subsamples were observed in the satisfaction of restraints. Structural differences among the variants, as well as between AlphaFold2 models and previously published comparative models of nanobodies, lead to differences in binding modes of the same nanobody to different spike variants. Thus, for the sake of consistency we limit our comparison to the receptor epitopes, which are defined as all receptor atoms that are within 6 Å of the framework and CDR regions of the nanobodies (excluding the flexible N- and C-terminal regions). Although we don't include the nanobody paratopes in our analysis, we verified that all binding modes are primarily through CDR3, except for the CDR1 contribution to the binding of S1-1 and S1-RBD-15 to the RBD, for all variants. Relative differences in binding affinity (K_d) and neutralization potential (IC₅₀) between the original virus and other variants (delta and omicron) were projected onto the Wuhan epitopes to create the heatmaps in **Figures 1B** [↗](#) and **2B** [↗](#). Relative differences were reported as $-\log\left(\frac{K_d^{\text{variant}}}{K_d^{\text{wuhan}}}\right)$ or $-\log\left(\frac{IC_{50}^{\text{variant}}}{IC_{50}^{\text{wuhan}}}\right)$ and normalized to the range from 0-100.

Integrative models of nanobody epitopes on the spike protein were computed on the Wynton HPC cluster at UCSF. Receptor epitopes were visualized in UCSF ChimeraX (Pettersen, Goddard et al. 2021 [↗](#)). Files containing input data, scripts and output results are available at https://github.com/integrativemodeling/nbspike/tree/main/integrative_modeling_VOC [↗](#). Structure predictions using

ColabFold utilized the “AlphaFold2_batch” notebook, with the default settings. All modeled structures were subjected to molecular-mechanics-based relaxation, followed by using the model with the top pLDDT score was selected for the integrative modeling pipeline.

Cell lines

TMPRSS2-expressing Vero E6 cells, 293T/17 cells and 293T-hACE2 cells were cultured as described previously (Mast, Fridy et al. 2021 [↗](#)). Briefly, TMPRSS2+ Vero E6 cells were cultured at 37°C in the presence of 5% CO₂ in medium composed of in high-glucose Dulbecco’s modified Eagle’s medium (DMEM, Gibco) supplemented with 10% (v/v) FBS and 1 mg/ml geneticin. 293T/17 were cultured at 37°C in the presence of 5% CO₂ in a medium composed of DMEM supplemented with 10% (v/v) FBS and penicillin/streptomycin. 293T-hACE2 cells were cultured at 37°C in the presence of 5% CO₂ in medium composed of DMEM supplemented with 10% (v/v) FBS, penicillin/streptomycin, 10 mM HEPES, and with 0.1 mM MEM non-essential amino acids (Thermo Fisher). All experiments were performed with cells passaged less than 15 times. The identities of cell lines were confirmed by chromosomal marker analysis and tested negative for mycoplasma using a MycoStrip (InvivoGen).

Production of SARS-CoV-2 variant pseudotyped lentiviral reporter particles

Pseudovirus stocks were prepared and 12ittered as described previously (Mast, Fridy et al. 2021 [↗](#)). Variant spike containing plasmids were combined with pHAGE-CMV-Luc2-IRES-ZsGreen-W (BEI Cat # NR-52516) (Crawford, Eguia et al. 2020 [↗](#)), and psPAX using lipofectamine 3000 and cotransfected into 293T/17 cells. Pseudovirus was 12ittered by threefold serial dilution on 293T-hACE2 cells, as described previously (Mast, Fridy et al. 2021 [↗](#)).

SARS-CoV-2 pseudovirus neutralization assay

Nanobodies were tested for their neutralization properties as described previously (Mast, Fridy et al. 2021 [↗](#)). Briefly, threefold serial dilutions of nanobodies were incubated with pseudotyped SARS-CoV-2 for 1 h at 37°C. The nanobody-pseudovirus mixtures were then added in quadruplicate to 293T-hACE2 cells along with 2 µg/ml polybrene (Sigma). Cells were incubated at 37°C with 5% CO₂. Infected cells were processed between 52 and 60 hr by adding equal volume of Steady-Glo (Promega), and firefly luciferase signal was measured using the Biotek Model N4 with integration at 0.5 ms. Data were processed using Prism 7 (GraphPad), using four-parameter nonlinear regression (least-squares regression method without weighting). All nanobodies were tested at least two times and with more than one pseudovirus preparation.

Plaque reduction neutralization assay with SARS-CoV-2 BA.5

Briefly, 10 three-fold serial dilutions of representative nanobodies of select epitope groups in Opti-MEM (Gibco) were incubated with approximately 100-200 pfus of SARS-CoV-2 BA.5 for 1hr at RT. The nanobody/virus mixture was then added to a confluent monolayer of TMPRSS2+ Vero E6 cells in 12-well plates and incubated at RT for 90 minutes. One well was overlaid with virus only while another well was uninfected. Virus/nanobody mixture was removed, and the cell monolayer overlaid with a medium composed of 3% (w/v) carboxymethylcellulose and 4% (v/v) FBS in Opti-MEM. 96 hr post infection, the overlay was removed and cell monolayer was washed with DPBS (Gibco) before being fixed with 4% (w/v) paraformaldehyde in DPBS for 30 minutes. Fixative was removed and the cells were rinsed with DPBS before being stained with 1% (w/v) crystal violet in 20% (v/v) ethanol. Contrast was enhanced by washing with DPBS, and clear plaques representing individual viral infections were visualized as spots lacking crystal violet stain. Plaques were quantified and the ratio of plaques at each nanobody dilution to ‘virus only’ well was used to determine the IC₅₀s of each nanobody.

Nanobody synergy

Synergy experiments were performed as described previously, (Mast, Fridy et al. 2021 [↗](#)). Briefly, a robotic liquid handler was used to prepare 2D matrices of threefold serial dilutions of two nanobodies and then mix these combinations with different variant pseudotyped SARS-CoV-2 for one h. After incubation with the virus, the mixture was overlaid on a monolayer of 293-hACE2 cells and left to incubate for 56 h. Luminescence was quantified as described above. Data were processed using the Bivariate Response to Additive Interacting Doses (BRAID) model (Twarog, Stewart et al. 2016 [↗](#)) as implemented in the synergy software package for python (Wooten and Albert 2021 [↗](#)).

SARS-CoV-2 stocks and titers

All experimental work involving live SARS-CoV-2 was performed at Seattle Children's Research Institute (SCRI) in compliance with SCRI guidelines for BioSafety Level 3 (BSL-3) containment. SARS-CoV-2 isolate CGIDR_SARS2 omicron BA.5 was obtained from an infected individual. An initial inoculum was diluted in Opti-MEM (Gibco) at 1:1000, overlaid on a monolayer of Vero E6 and incubated for 90 min. Following the incubation, the supernatant was removed and replaced with 2% (v/v) FBS in Opti-MEM medium. The cultures were inspected for cytopathic effects, and infectious supernatants were collected, cleared of cellular debris by centrifugation, and stored at –80°C until use. Whole viral genome sequencing and variant analysis was performed by the University of Washington Department of Laboratory Medicine & Pathology. Viral titers were determined by plaque assay using a liquid overlay and fixation-staining method, as described previously (Mendoza, Manguiat et al. 2020 [↗](#), Mast, Fridy et al. 2021 [↗](#)).

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Competing interests

JDA, BTC, MPR, PCF, NEK, FDM, and JPO are inventors on a provisional patent describing the anti-spike nanobodies described in this manuscript.

Supplemental material

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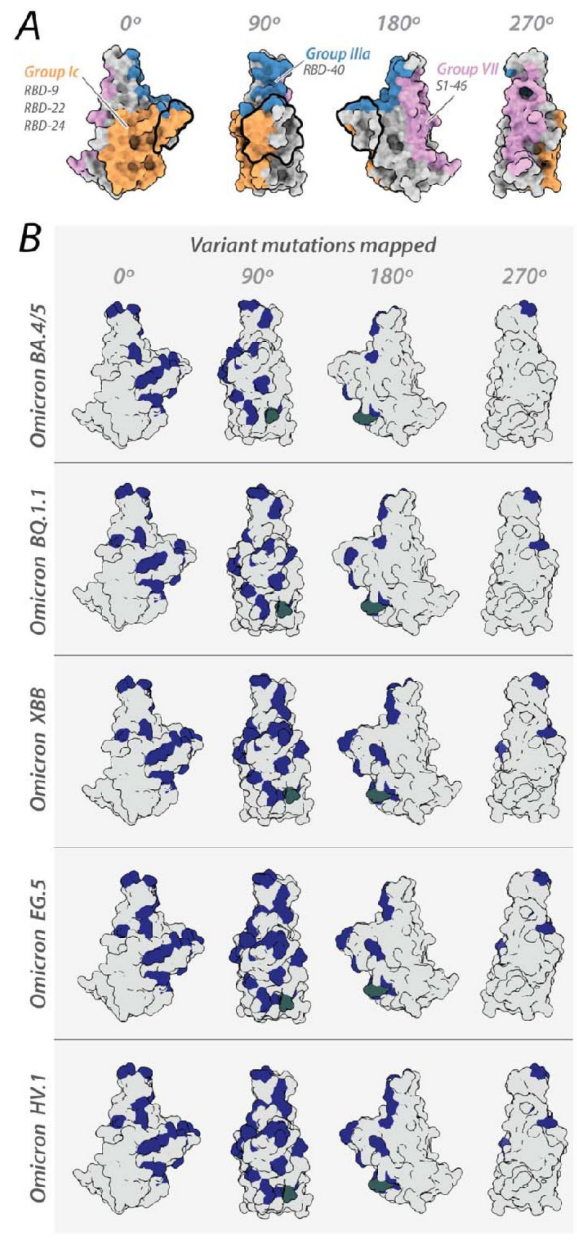


Figure 5–Figure supplement 1.

Mapping of variant specific amino acid changes on different SARS-COV-2 variants.

(A) Nanobody epitopes predicted to retain binding/neutralization to current circulating SARS-CoV-2 VoC mapped on wild-type Spike RBD (PDB ID: 6M0J). (B) The RBD of each variant of concern is colored gray and shown in four different orientations. Amino acid differences of each variant are colored blue. The mutations were mapped on the RBDs of spike omicron BA.4 RBD (modeled), omicron XBB.1, EG.5 and HV.1 (all three mapped on PDB ID: 8IOU) and omicron BQ.1.1 (PDB ID: 8FXC). All structure representations were created on ChimeraX (Pettersen, Goddard et al. 2021 [DOI](https://doi.org/10.1093/bioinformatics/btab351)).

Figure 4—Source data 1. Neutralization data from the PRNT assay with omicron BA.5 virus.

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Figure 6—Source data 2. Neutralization data from synergy experiment

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Reviewer #1 (Public Review):

Summary:

In this manuscript, Ketaren, Mast, Fridy et al. assessed the ability of a previously generated llama nanobody library (Mast, Fridy et al. 2021) to bind and neutralize SARS-CoV-2 delta and omicron variants. The authors identified multiple nanobodies that retain neutralizing and/or binding capacity against delta, BA.1 and BA.4/5. Nanobody epitope mapping on spike proteins using structural modeling revealed possible mechanisms of immune evasion by viral variants as well as mechanisms of cross-variant neutralization by nanobodies. The authors additionally identified two nanobody pairs involving non-neutralizing nanobodies that exhibited synergy in neutralization against the delta variant. These results enabled the refinement of target epitopes of the nanobody repertoire and the discovery of several pan-variant nanobodies for further preclinical development.

Strengths:

Overall, this study is well executed and provides a valuable framework for assessing the impact of emerging SARS-CoV-2 variants on nanobodies using a combination of in vitro biochemical and cellular assays as well as computational approaches. There are interesting insights generated from the epitope mapping analyses, which offer possible explanations for how delta and omicron variants escape nanobody responses, as well as how some nanobodies exhibit cross-variant neutralization capacity. These analyses laid out a clear path forward for optimizing these promising next-gen therapeutics, particularly in the face of rapidly emerging SARS-CoV-2 variants. This work will be of interest to researchers in the fields of antibody/nanobody engineering, SARS-CoV-2 therapeutics, and host-virus interaction.

Weaknesses:

A main weakness of the study is that the efficacy statement is not thoroughly supported. While the authors comprehensively characterized the neutralizing ability of nanobodies in vitro, there is no animal data involving mice or hamsters to demonstrate the real protective efficacy in vivo. Yet, in the title and throughout the manuscript, the authors repeatedly used phrases like "retains efficacy" or "remains efficacious" to describe the nanobodies' neutralization or binding capacities. This claim is not well supported by the data and underestimates the impact of variants on the nanobodies, especially the omicron sublineages. For example, the authors showed that S1-RBD-15 had a ~100-fold reduction in neutralization titer against Omicron, with an IC50 at around 1 μ M. This is much higher than the IC50 value of a typical anti-ancestral RBD nanobody reported in the previous study (Mast, Fridy et al. 2021). In fact, the authors themselves ascribe nanobodies with an IC50 above 1 μ M as weak neutralizers. And there were many in the range of 0.1-1 μ M. Furthermore, many nanobodies selected for affinity measurement against BA.4/5 had no detectable binding. Without providing in vivo protection data or including monoclonal antibodies that are known to be efficacious against variants in the in vitro assays as a benchmark, it is difficult to evaluate the efficacy just with the IC50 values.

Comments post revision:

The authors are to be commended for their comprehensive response to the referees' comments. In the revised manuscript, the authors made extensive changes throughout the texts and added new figures that greatly improved their clarity. While the manuscript is still limited in solely relying on in vitro data for efficacy assessment, it nicely demonstrates how the combination of experimental and computational techniques could lead to the discovery of broadly neutralizing nanobody candidates for further lead optimization.

<https://doi.org/10.7554/eLife.89423.2.sa1>

Reviewer #2 (Public Review):

Summary:

Interest in using nanobodies for therapeutic interventions in infectious diseases is growing due to their ability to bind hidden or cryptic epitopes that are inaccessible to conventional immunoglobulins. In the presented study, authors posed to characterize nanobodies derived the library produced earlier with Wuhan strain of SARS-CoV-2, map their epitopes on SARS-CoV-2 spike protein and demonstrate that some nanobodies retain binding and even neutralization against antigenically distant, newly emerging Variants of Concern (VOCs).

Strengths:

Authors demonstrate that some nanobodies despite being obtained against ancestral virus strain retain high affinity binding to antigenically distant SARS-CoV-2 strains despite majority of the repertoire loses binding. Despite being limited to only two nanobody combinations, demonstration of synergy in virus neutralization between nanobodies targeting different epitopes is compelling. The ability of nanobodies to bind emerging virus strains has been demonstrated and the possible effect of mutations within epitopes has been thoroughly discussed.

<https://doi.org/10.7554/eLife.89423.2.sa0>

Author response:

The following is the authors' response to the original reviews.

Reviewer #1 (Recommendations For The Authors):

Beyond my general review, some descriptions of the results and methods could be further clarified, which I've outlined below:

(1) Page 3, Line 118-120: Based on results from Fig 1A, the authors reported 15 nanobodies neutralized both delta and BA.1 out of the 41 tested. However, I only counted 14. Could the authors double check?

We recounted the nanobodies and confirmed there are 15 as follows:

- (1) RBD-15
- (2) RBD-22
- (3) RBD-24
- (4) RBD-9S1-4
- (5) S1-35
- (6) RBD-6
- (7) RBD-5
- (8) RBD-21
- (9) RBD-16
- (10) S1-46
- (11) S1-49dimer
- (12) S2-10dimer
- (13) S2-3
- (14) S2-62

(2) Page 5, Lines 134-135: the authors described that the heatmap reflects the neutralizing strength of the representative nanobodies from each group. For groups where multiple nanobodies were selected for visualization, how was the neutralization strength calculated? Was the IC50 averaged first before being converted into the neutralization strength?

This has been made clear in the legend for Fig. 1 as follows “For groups with multiple nanobodies, the average $-\log_{10}$ (IC50) is first calculated for the nanobodies within that group, then normalized to a neutralization score within the 0–100 range using the min and max average $-\log_{10}$ (IC50) for that group. A higher score indicates more potent neutralization of the variant relative to the wild type.”

(3) Page 5, Lines 138-139: What was the authors' rationale for selecting certain nanobodies over others for structural modeling and visualizing the neutralization

heatmap in Fig 1B? Does it introduce bias to the neutralizing epitope map on the spike protein?

We only focused on nanobodies for which we had enough epitope mapping data to unambiguously generate docked nanobody-spike models, as explained in our previous study (Mast et al, eLife 2021). When multiple nanobodies within the same group had sufficient epitope mapping data available, we selected only representative candidates that had better binding affinity and/or neutralization potency. As epitope mapping via escape mutants relied largely on random point mutagenesis of Spike, there should be little introduced bias.

Overall, groups I-VII cover an exhaustive set of target areas on the RBD (including the lone glycan site in Group-II), while groups VIII and IX are representative areas on NTD and S2. Using group-average IC50s and suitable normalization as mentioned in point 3 above further prevent potential biases due to unequal number of Nbs modeled from each group.

We have modified the text with the following:

“For computational epitope modeling, we selected nanobody candidates using a series of experimentally obtained structural restraints, as described in Mast, Fridy et al. 2021.”

(4) Page 5, Lines 161-167: It would be good to include Fig S1 as a main figure as it places the epitope landscape of nanobodies being investigated in this manuscript into the broader context of clinically approved monoclonal antibody therapeutics for COVID-19.

We have amended the Figures to accommodate the reviewers suggestion. Figure S1 is now Figure 2.

(5) Page 6, Lines 173-175: The neutralization breadth for S1-46 is quite encouraging. Any speculations on why this particular nanobody is so broadly targeting? Any additional thoughts on why its high binding affinity (nM) did not translate into strong neutralization (as it is in the 0.1-1 uM range)?

S1-46 binds a region on spike that is conserved across all variants observed to date. Its epitope is difficult to access unless the RBD is in the up conformation, which may explain why monoclonal antibodies rarely bind. We state this in the text as follows:

“S1-46 binds a region on spike that is conserved across all variants to date, but which may be relatively inaccessible and is not targeted by any of the mAbs that previously received EUA by the FDA (Cox, Peacock et al. 2023).”

Relating neutralization activity to binding activity requires more insight into the mechanisms of binding and activity. Nonetheless, we are also encouraged by S1-46’s breadth and numerous avenues can be pursued to greatly improve its neutralizing activity (e.g. synergistic combinations).

(6) Page 6, Lines 173-175: For the remaining two nanobodies S1-31 and S1-RBD-11 in group VII, the target epitopes on the spike proteins of either delta or BA.1 do not seem to bear any mutations, at least based on the mutation maps in Fig 1B. Yet their neutralizing capacities against delta and BA.1 variants were abolished. Do the authors have any idea about what is going on here?

For group VII, only the epitope of S1-46 was mapped whereas S1-31 and S1-RBD-11 were assigned to group VII based on our lower resolution binning experiments. Thus, without knowing precisely where they bind, we can make only limited conclusions at this time. In the absence of supporting structural information, we speculate that the epitopes of RBD-11 and

S1-31 may be in a region that overlaps with or is in close proximity to a mutation that could affect the binding of the nanobody enough to result in loss of neutralizing ability.

(7) Page 7, Line 195-200: Please provide PRNT50 or logPRNT50 for the five nanobodies selected for BA.4/5 PRNT assay.

We have added this suggested information. Additionally, a supporting table (Table S1) is now provided.

(8) Page 8, Lines 223-224: Similar to comment 3, what was the rationale here for choosing certain nanobodies over others for structural modeling and visualizing the binding heatmap in Fig 2B?

The set of nanobodies chosen for structural modeling and visualization of neutralization data is identical to the set of anti-RBD nanobodies chosen for binding.

(9) Page 11, Lines 326-328: Can the authors include mutation maps as part of Fig 4C to show the mutation distributions on the XBB/BQ.1/BQ.1.1 spikes?

We have updated and added a supplemental figure to accompany Fig. 5 (called “supplement for Figure 5”) showing the mutation maps.

(10) Page 14, Line 409-418: This paragraph is well considered. Given the large number of nanobodies assessed in this manuscript, it would be helpful if the authors could highlight some candidate nanobodies as lead candidates for further optimization.

While our intention in this manuscript was not to provide targeted recommendations for lead candidates, but rather to reiterate the collective potential of a Nb pool originally targeted towards the 2019 Wuhan variant, the reviewers point is interesting. We speculate that any of the Nbs we have demonstrated to show pan-VoC activity, would be prime candidates for further optimization.

We have added a statement to this effect as follows: “We propose that any of the Nbs we have demonstrated to show pan-VoC activity, would be prime candidates for further optimization.”

Reviewer #2 (Recommendations For The Authors):

Major concerns:

(1) The main message of the article is the prediction that nanobodies that retain binding to the different SARS-CoV-2 variants including early Omicron strains will retain binding and neutralization against currently circulating strains such XBB and BQ. However, no evidence either via modeling or experimental testing has been provided for that prediction. The study will benefit from mapping amino acid mutations in RBD of XBB and BQ lineages compared to BA.4/5 and demonstrating via computation docking that epitopes of the five nanobodies that retain binding to BA.4/5 RBD are not affected. For example, the crystal structure of XBB.1 RBD PDB:8OIV is available. Binding/neutralization experiment with currently circulating SARS-CoV-2 strains would still be the gold standard test given the fact that only five out of 41 nanobodies retained binding and neutralization to BA.4/5 lineage. Loss of neutralization ability against BA.4/5 without a significant decrease in binding affinity for nanobodies S1-46 and S1-RBD-22 further indicates that neutralization of XBB and BQ lineage should be performed.

The docking protocol used to predict the spike epitopes uses a C-alpha resolution to represent protein residues, and is data-driven, i.e. it assumes that binding happens in the first place,

and then utilizes experimentally obtained structural restraints. So, concluding possible binding from such a docking protocol alone would be noisy. In our revised manuscript we have a new Figure 3B, which shows epitopes of 4 out of the 5 pan-VoC nanobodies, i.e. S1-RBD-{9, 22, 40) and S1-46 mapped to the RBD structures of XBB.1 (8IOU) and BQ.1.1 (8FXC), and we have updated Figure 4 with a supplemental showing the mutation maps.

(2) Described nanobodies are positioned as very potent neutralizers of SARS-CoV-2. However, they are much less potent in neutralization of ancestral strain as well as early VOCs compared to the mAbs that were approved for COVID-19 treatment. For example, IC50 for casirivimab and imdevimab are 37.4 pM and 42.1 pM, respectively. That is about 27-fold more than IC50 for the most potent nanobody reported in the article, S1-RDB-15.

This comparison is fraught for several reasons. 1. Experimental differences in pseudovirus assay systems usually result in significant differences in reported IC50s, as IC50 is not an absolute measure, or ultimately comparable to clinical IC50 values. For this reason, in our original publication (Mast et al., 2021) we tested other nanobodies in our experimental set-up as benchmarks (Mast et al., 2021). 2. A typical monoclonal has two binding sites with a large structural Fc linker that is combined ~10 times the size of a nanobody. In a therapeutic setting where monoclonal therapy is provided in g per kg of patient body weight, there is a 5-fold excess of Nb binding to antibody binding capacity. 3. We have previously shown that dimerizing our nanobodies (to produce two antigen binding sites) can dramatically increase potency over 100 fold (Mast et al., 2021).

In order to make this even clearer in the manuscript, we have added the following: “We note that IC50s are not directly comparable across different experimental set-ups because measured values are highly dependent on the experimental conditions. For this reason, we included other published nanobodies as benchmarks in our original publication and have subsequently maintained standard experimental conditions (Mast, Fridy et al. 2021)”.

(3) Figure 1A. If each dot represents an independent measurement of the same nanobody, IC50 variation seems too high. For some nanobodies it ranges for almost a log of magnitude, e.g S1-RDB-24, S1-RBD-46, S2-3. Why is that?

We have deliberately explored the full range of effects that could contribute to experimental variability in our pseudovirus assay, using different batches of nanobody and pseudovirus in each replicate to provide as impartial and comprehensive analysis as possible. While the activity of some nanobodies is remarkably stable from batch to batch, others show the variation noticed by the Reviewer, hence why we performed multiple replicates to define the average IC50 value for our nanobodies.

(4) The drop in IC50 for BA.1 neutralization is about one log for the majority of tested nanobodies. This should be outlined in the text. For example, for the most potent neutralizer, S1-RDB-15, the drop in IC50 for BA.1 is about 100-fold compared to IC50 for the Delta and Wuhan strains. It is important to note that out of 9 nanobodies for that drop in neutralizing capacity against BA.1 and Delta variants less than one log of magnitude 2 have epitopes in the S2 domain of SRS-CoV-2 spike. Resistance of mAbs targeting the S2 part of the spike has been extensively described in the literature as being due to the highly conserved structure of this region that facilitates membrane fusion. Presented data demonstrate that >80% of the nanobody repertoire is affected by mutations on spike protein. Additionally, it can be helpful for readers if the fold-change in IC50 between Wuhan, Delta, and BA.1 is presented in the text or added to Figure 1 or a table.

We agree with the Reviewer and to make this more explicit we have made the following change: “In comparison, groups I, I/II, I/IV, V, VII, VIII and the anti-S2 nanobodies contained the majority of omicron BA.1 neutralizers, though here the neutralization potency of many nanobodies was generally decreased tenfold compared to wild-type (emphasis added).”

(5) The authors should either present the results of the formal correlation analysis or avoid using misleading verbiage such as: "the decrease in neutralization potency largely correlates with the accumulation of omicron BA.1 specific mutations throughout the RBD" or "significant decrease in binding affinity correlated to decreases neutralization potency".

We thank the Reviewer for this constructive feedback. To address this question, we have performed a correlation analysis using Pearson and Spearman's methods to quantitatively assess the relationship between nanobody neutralization potency (IC50) and binding affinity (KD) across SARS-CoV-2 variants, including the wildtype, delta, and omicron BA.1 variants. Our results indicate a statistically significant correlation for the delta variant (Pearson's PCC: 0.71, p-value: 0.01; Spearman's rho: 0.63, p-value: 0.07), supporting our statement regarding the correlation between decreased neutralization potency and reduced binding affinity for this variant. However, for the wildtype and omicron BA.1 variants, the correlations were not statistically significant (wildtype Pearson's: 0.10, p-value: 0.70; omicron BA.1 Pearson's: 0.27, p-value: 0.31), which we acknowledge does not fully align with the verbiage used in the manuscript. Therefore, we have revised the manuscript to present the correlation analysis data accurately and ensure the discussion is reflective of the statistical evidence as follows:

“SPR binding assessments to the spike S1 domain or RBD of delta revealed a pattern: nanobodies maintaining binding affinity generally also neutralized the virus with a statistically significant correlation between binding affinity and neutralization efficacy (Pearson's Correlation Coefficient: 0.71, p-value: 0.01; Spearman's rho: 0.63, p-value: 0.07). However, this correlation was not statistically significant for omicron BA.1 (Pearson's Correlation Coefficient: 0.27, p-value: 0.31) (Fig. 3A, Table 1). Notably, while some nanobodies bound to the variants, they did not consistently neutralize them, suggesting additional factors influence neutralization beyond mere binding.”

(6) Figure 3 shows approximated curves for live virus neutralization assay with quite a broad 90% CI. It will be helpful to present, at least, in supplementary, primary data for live-virus neutralization that were used to perform non-linear regression.

We have added the reviewer's suggestion.

(7) It is not clear what are the "variant-specific nanobody groups" exactly? A definition/description of the term is not provided. If the nanobody library was generated with the Wuhan strain, how did strain-specific nanobodies that bind/neutralize only Delta, BA.1 or BA.4/5 appear in the repertoire and were isolated? This statement also contradicts data in Table 4 where all nanobodies listed bind and neutralize Wuhan strain.

We agree with the reviewer. All nanobodies tested bind/neutralize the Wuhan strain as they were selected from our original repertoire of 116 nanobodies (Mast, et al., 2021). To clarify, variant-specific nanobodies are nanobodies that bind only one variant that arose from the original Wuhan strain. They were categorized into variant-specific groups based on whether they were able to bind each variant (other than Wuhan).

We have thus added to the manuscript, “we define variant-specific nanobodies as nanobodies that bind a single additional variant alongside the original Wuhan strain...”

(8) Describing the categorization of nanobody epitope groups presented in Figure 4, the authors state that binding to Wuhan, Delta, BA/1, and BA.4/5 predicts that these nanobodies will be "effective binders against current circulating strains of the virus including XBB and BQ lineages"? How exactly is this conclusion corollary to the data shown?

The epitopes of XBB and BQ.1 are not divergent enough within the regions we propose the nanobodies to bind, to suggest that nanobodies that bind in those regions will lose binding ability. We hypothesize that the region at which these nanobodies bind represents regions on spike that are vulnerable to our specified nanobodies in Fig. 4. We have generated a new Fig. 3B and added a supporting figure for Fig. 4 to address this.

(9) Figures 4C and 6 describe how the nanobodies will retain binding to currently circulating strains of XBB lineage. However, epitopes are mapped on the same Wuhan, Delta, BA.1, and BA.4/5 virus strains. The predicted binding of nanobodies to XBB lineage RBD is not actually shown in Figure 6. It is clear from the figure that the nanobody binding footprint (red area) decreases with antigenic distance in every spike projection from Wuhan through the BA.4/5 strain. It is unclear how this indicates that nanobodies will remain active against even more distant XBB, BQ, EU, and CH strains accumulating more mutations in spike protein.

We have added the following to the manuscript to clarify: “Strikingly, we have in our cohort 8 nanobodies able to bind delta, and the omicron lineages BA.1/BA.4/BA.5/XBB/BQ.1.1 (Fig. 5B). We further predict these 8 nanobodies will be effective binders against current circulating strains of the virus including omicron EG.5 and HV.1 as the epitope regions (or predicted epitopes) of these nanobodies do not vary significantly from omicron lineages XBB and BQ.1.1 (Fig. 5C and Supplement to Fig. 5).”

(10) Despite major advances in the development of nanobodies as therapeutic molecules there are only a few nanobody-based drugs that have so far been approved for clinical use and all of them are nanobody fusions to immunoglobulin Fc fragment. It is dictated by the small size of the nanobody itself, 15 kDa molecule, that leads to rapid kidney clearance within hours post-injection, and also by the necessity of having antibody effector functions allowing for example killing of malignant cells. It is hard to predict how each individual nanobody will tolerate multimerization and if it will still retain binding ability as its size dramatically increases. It should be noted that IC50 for BA.4/5 is in the submicromolar range for the 5 nanobodies retaining neutralization of this strain. From a therapeutic perspective, this is quite a high IC50 that dictates a high dosage to achieve a therapeutic effect. Furthermore, it can be expected that additional mutations in the SARS-CoV-2 spike will further affect binding affinity and therefore reduce the neutralization ability of these nanobodies resulting in even higher doses required to achieve therapeutic effect. Therefore, authors should discuss the limitations of the nanobody approach as a therapeutic intervention more granularly.

While Fc fusions are not strictly required for clinical use (for instance Caplacizumab is not an Fc fusion, being a multimer containing an albumin-binding nanobody), we agree that reformulation would indeed be required to optimize pharmacokinetics for eventual clinical use. Increased valency through multimerization is in fact one of several strategies, which also includes synergistic combinations, for significantly enhancing effective IC50. Preclinical nanobody engineering is not within the scope of this paper, but we acknowledge this challenge.

Minor points:

| (1) *Table S1 is missing.*

This is an .xlsx file uploaded as Supplementary File 3. Labeled now as “Figure 6–Source data 2. Neutralization data from synergy experiment”.

| (2) *Because Table 1 summarizes all neutralization and binding data, it will be helpful to refer to it while describing data presented in Figure 1.*

This has been added to the revised manuscript.

| (3) *Live SARS-CoV-2 PRNT is not described in Materials and Methods.*

This has been added to the revised manuscript.