

## PROTEIN DESIGN

# Target-conditioned diffusion generates potent TNFR superfamily antagonists and agonists

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Despite progress in designing protein-binding proteins, the shape matching of designs to targets is lower than in many native protein complexes, and design efforts have failed for the tumor necrosis factor receptor 1 (TNFR1) and other protein targets with relatively flat and polar surfaces. We hypothesized that free diffusion from random noise could generate shape-matched binders for challenging targets and tested this approach on TNFR1. We obtain designs with low picomolar affinity whose specificity can be completely switched to other family members using partial diffusion. Designs function as antagonists or as superagonists when presented at higher valency for OX40 and 4-1BB. The ability to design high-affinity and high-specificity antagonists and agonists for pharmacologically important targets in silico presages a coming era in protein design in which binders are made by computation rather than immunization or random screening approaches.

The design of proteins that bind with high affinity and specificity to targets of interest is a long-standing challenge in computational structural biology with applications in therapeutics, diagnostics, and beyond (1, 2). To address this problem, protein design methods have generally relied on preexisting sets of scaffolds—either native proteins or de novo designs—with well-defined tertiary structures (3–6). For example, binders were generated for a number of targets by docking a large set of idealized ~65-residue protein scaffolds and identifying those docks capable of hosting the lowest-energy binding interactions (3). The generative artificial intelligence method RFdiffusion generates new protein structures by progressively denoising random spatial distributions of residues; this diffusive denoising process can be guided by conditioning information to create a wide range of protein structures and functions—for example, to design binding proteins, diffusion trajectories are carried out in the presence of the fixed target of interest (7). RFdiffusion has previously been used to generate binders conditioned on similar ideal scaffolds; unconstrained diffusion was also tested but only yielded simple three- or four-helix bundles (7). The advantage of using small ideal scaffolds with regular secondary-structure elements and packing is that after sequence design, a reasonable fraction of designs are likely to fold as expected, but this limits the extent of shape matching achievable, particularly for targets with relatively flat surfaces lacking concavities for small mini-proteins to fit into. Indeed, the contact molecu-

lar surface (CMS) of de novo–designed binding proteins to date is lower than that of many native protein complexes (Fig. 1A).

We reasoned that the limited shape matching achievable by scaffold-dependent and/or short-chain approaches could be overcome by using RFdiffusion to directly generate larger proteins starting from completely random residue distributions in the presence of the target of interest without any guidance from preexisting scaffolds. Completely unconstrained RFdiffusion can generate a wide diversity of folds and assemblies (7), and proteins with folds that wrap around extended helical peptides (8). We set out to explore whether RFdiffusion trajectories conditioned only on the structure of a folded protein target, and not biased toward any particular scaffold or limited by the number of available residues, could generate folds with extensive shape matching to the target and enable the design of high-affinity binders to targets for which previous design efforts had failed.

## Binder design

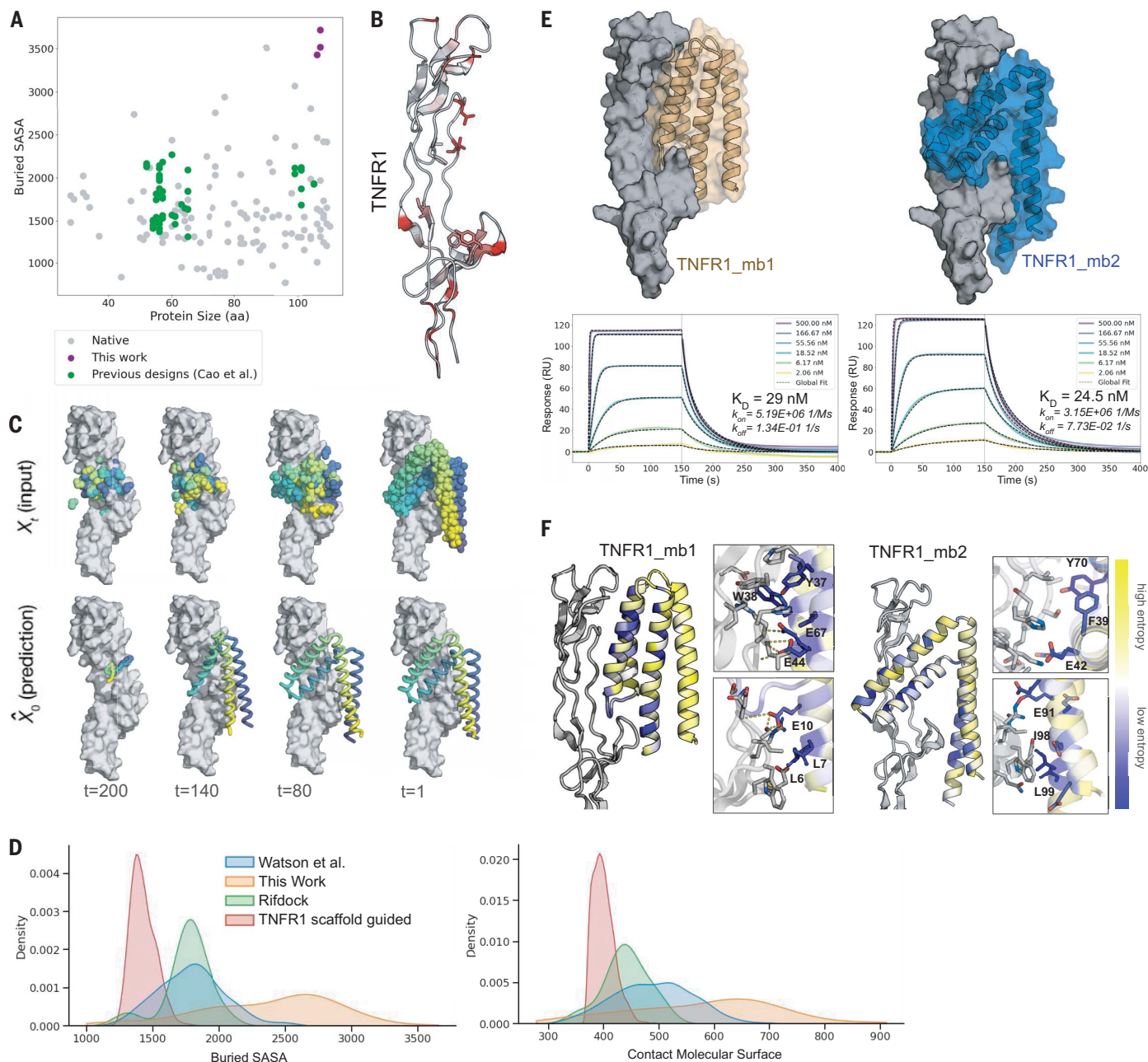
We chose to focus on the tumor necrosis factor receptor superfamily (TNFRSF), which includes many important drug targets, including the TNF receptor 1 (TNFR1), which plays a key role in inflammatory disease (9). Like other family members, TNFR has an extended flat and largely polar surface lacking the concave sites with somewhat hydrophobic sites that previous de novo design efforts have successfully targeted (Fig. 1B). Indeed, multiple attempts to generate binders to TNFR and other family members using the approach of Cao *et al.* (3) met with little success. We first attempted to use RFdiffusion as described in Watson *et al.* (7) to generate binders to TNFR1, using guidance from 65–amino acid scaffold libraries or limited to chains less than 65 residues but also had little success (fig. S1).

We next set out to adapt RFdiffusion to generate backbones with a more extended contact surface that can engage dispersed surface hydrophobic residues (Fig. 1B). Protein interactions, such as protein folding, are largely driven by hydrophobic interactions, and targets with only a few surface hydrophobic residues distant in space have been particularly challenging. On TNFR, the few surface hydrophobic residues are separated by distances of up to 28 Å—too small for 65-residue proteins to simultaneously engage. Random Gaussian residue clouds of up to 120 residues were placed at the native ligand (TNF) interface, and RFdiffusion was biased to form contacts with these dispersed hydrophobic residues. This step generated a variety of backbones complementary in shape to TNFR1 (fig. S2) and quite different from those in previous campaigns. ProteinMPNN was used to design sequences for these backbones in complex with TNFR1 to favor both folding to the intended structure and binding to the target. To further sample around promising designs predicted by AlphaFold2 (AF2) to form complexes, as defined by a low predicted aligned error (pae) in the interface region (pae\_interaction <20), we extracted additional backbones from preceding diffusion timesteps, as the predicted structure was already close to the final structure at around timestep 120 out of 200 (Fig. 1C). The designs predicted to most strongly bind the TNFR1 (interface pae <7.5) and fold to the target backbone (predicted local distance difference test, pLDDT >85) were selected for experimental characterization (table S1). The contact molecular surface (CMS) and buried solvent-accessible surface area (SASA) of these designs were substantially higher than those of designs from previous campaigns (Fig. 1D).

Genes encoding 96 designs were obtained, and the proteins were expressed in *Escherichia coli*. Despite having less regular structures and longer lengths than most previous binder designs, 90 of 96 designs expressed well and were primarily monomeric (fig. S3A). Six of the designs bound to TNFR1 in surface plasmon resonance (SPR) experiments (fig. S3B), with designs TNFR1\_mb1 and TNFR1\_mb2 having dissociation constants ( $K_D$ 's) of 29 and 24.5 nM, respectively. Both designs were highly specific, with no detectable binding to TNFR2. The tertiary structure was quite distinct from that of previously designed de novo binders, and the designs interact with TNFR1 over an extended region (Fig. 1D). Design TNFR1\_mb2 has an unusual V-shaped fold with a very high CMS for TNFR of 795 Å<sup>2</sup>, substantially more than the average 490 Å<sup>2</sup> of previous diffused designs (7). Design TNFR1\_mb1 also has a binding mode less regular than that of previous de novo–designed binders with a connecting loop inserted in the TNFR1 binding cleft, and a

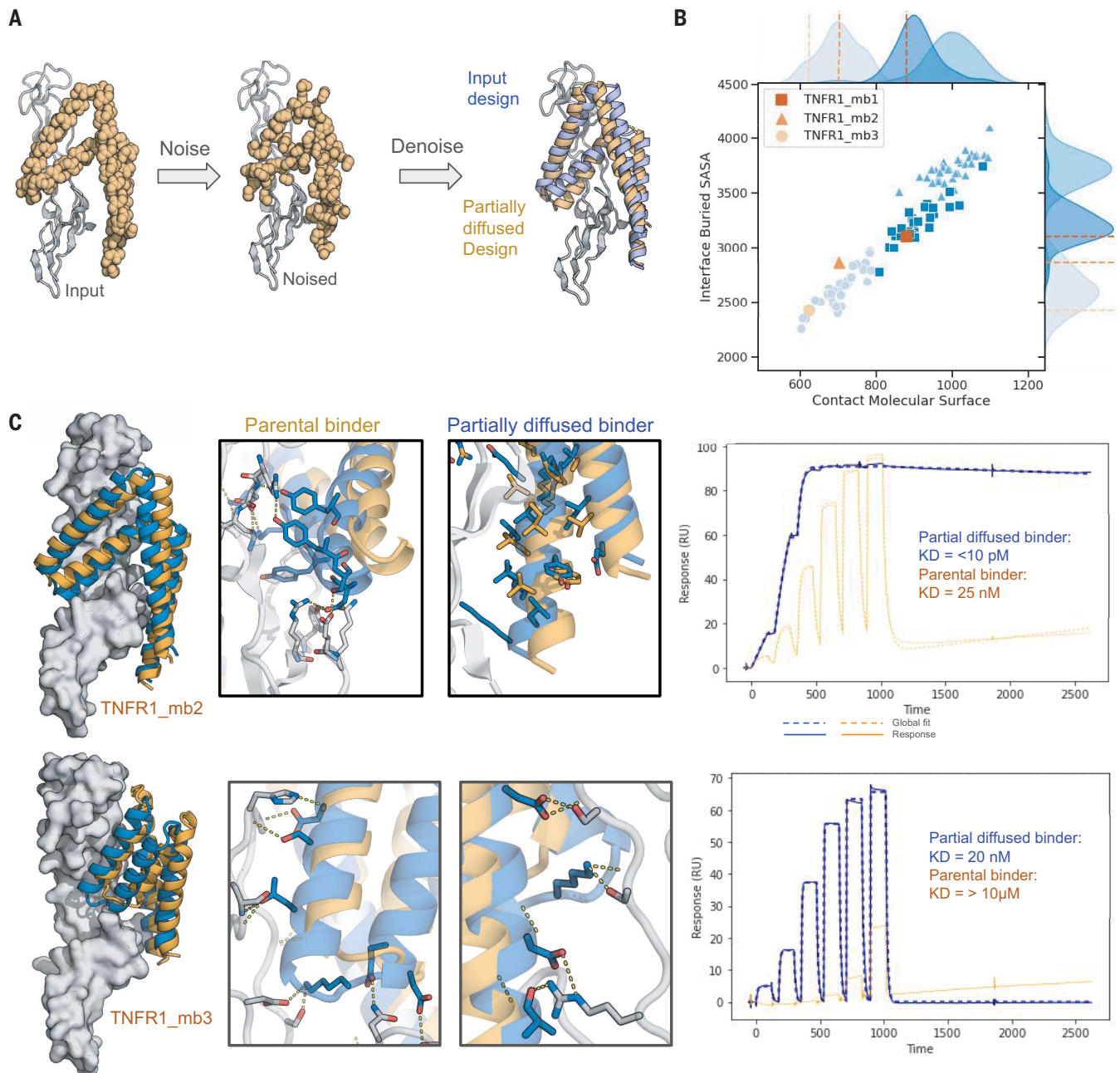
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**Fig. 1. Diffusion of shape-complementary binding proteins.** (A) Previous de novo–designed binders (green) generated by docking pregenerated scaffolds bury less surface area ( $\text{\AA}^2$ ) against their target than many native complexes (gray); the approach developed here enables the design of very large interfaces (purple). (B) TNFR1 is a challenging target, with a flat surface and few surface hydrophobic residues, which are shown in red on the structure (PDB ID: 6KP8). Residues selected as target hotspots for RFdiffusion are shown as sticks. (C) Representative RFdiffusion trajectory against TNFR1 starting from a random residue distribution placed against the target (top left). At each denoising step ( $X_t$ , top), the network generates a predicted structure ( $\hat{X}_0$ , bottom) and interpolates toward this structure to generate the next step ( $X_{t-1}$ ). (D) Comparison with previous TNFR1 design efforts. The RFdiffusion TNFR1 binder designs generated here (orange) have substantially higher buried SASA and CMS than designs against TNFR1 generated previously using the Rosetta RIF dock method (red) that failed

to bind. Designs generated against multiple targets using RIFdock (3) (green) and short chain RFdiffusion (7) (blue) are also shown for comparison. (E) (Top) Design models of binders TNFR1\_mb1 (orange) and TNFR1\_mb2 (blue) in complex with TNFR1 (gray). (Bottom) SPR measurements of binding to TNFR1. (F) SSM results confirm design models of TNFR1 binders and associated entropy. All 2014 and 2033 single–amino acid substitutions for TNFR1\_mb1 and TNFR1\_mb2 were expressed on yeast surface and probed using fluorescence-activated cell sorting with biotinylated TNFR1 followed by deep sequencing. Positions that were strongly conserved (low entropy, blue) were in the core and at the binding interface, whereas most surface residues away from the interface had high entropy (yellow). Entropy was calculated on the basis of the overall change of affinity (table S2). Zoom-ins show dense interaction networks of low-entropy residues (TNFR1 is in gray). Amino acid residues: E, Glu; F, Phe; I, Ile; L, Leu; W, Trp; and Y, Tyr.



**Fig. 2. Partial diffusion generates picomolar binders.** (A) Schematic of partial diffusion process for TNFR1\_mb2. The backbone of the input structure is represented as a collection of independent residues (first panel), noise is added (second panel), and then RFDiffusion is used to remove the noise, which results in a similar but better-fitting model (orange, right) compared to the input structure (purple). (B) Partial diffusion increases interface contacts. Contact molecular surface and interface buried solvent-accessible surface area are depicted for input designs

(orange) and the respective partial diffused variants in blue squares (TNFR1\_mb1), triangles (TNFR1\_mb2), and circles (TNFR1\_mb3). (C) Partial diffusion increases interface interaction density and binding affinity. For TNFR1\_mb2, an additional interface forms (left panel), whereas an existing interface remains largely unchanged (right). For TNFR1\_mb3, improved shape matching leads to additional interactions (bottom middle insets). The corresponding TNFR1 SPR traces are on the right (six steps of 5× dilutions from 500 nM).

similarly high CMS of 897 Å<sup>2</sup>. For both designs, we obtained a high-resolution binding footprint by determining the effect on binding of every amino acid substitution at every position one at a time (4047 substitutions in total) (figs. S4 and S5). The results were closely consistent with the design models, with substitutions affecting

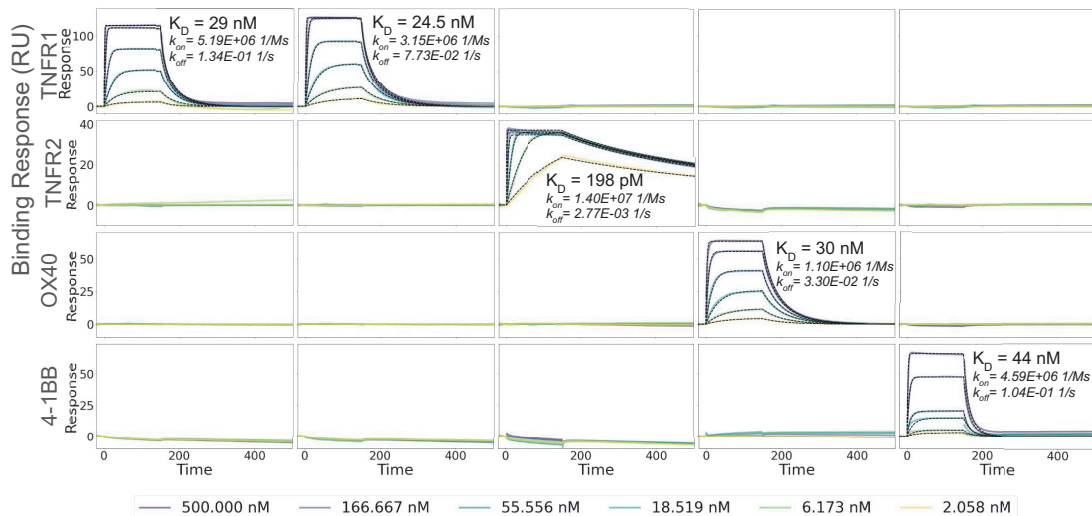
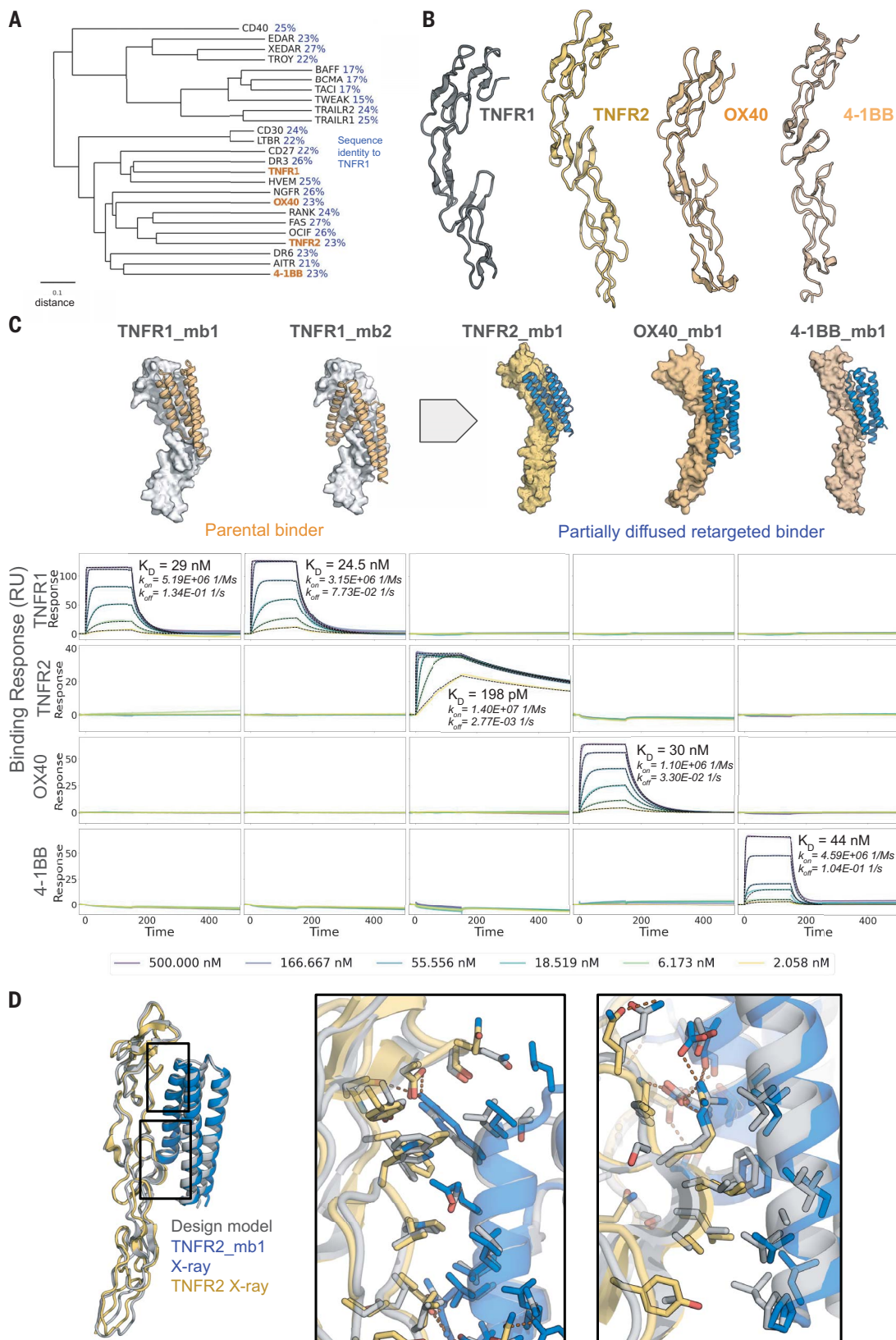
binding concentrated at the designed interface and in the protein core where they would disrupt folding. The most conserved interactions for both designs centered around the hydrophobic patches involving TNFR1 residues 107 and 111 and residues 38 and 40 (Fig. 1E and table S2) spanned by more polar interactions in the

center of the interface. Taken together, the close agreement of the site saturation mutagenesis (SSM) footprints with the design model interfaces, and AF2 and RoseTTAFold2 (RF2) complex predictions with the designed structures [pAE interface of 5.1 and 4.1, CA root mean square deviation (RMSD) to design of 0.6 and

### Fig. 3. Partial diffusion generates high-specificity TNFR2, OX40, and 4-1BB binders.

**(A)** Phylogenetic tree of TNFR superfamily constructed using the unweighted pair group method with arithmetic mean (UPGMA) based on a multiple sequence alignment performed with ClustalW. TNFR1, TNFR2, OX40, and 4-1BB investigated in this study are highlighted in orange. The scale bar represents a distance of 0.1 substitutions per site. Pairwise sequence identity to TNFR1 is indicated in blue. **(B)** Comparison of structures of TNFR1, TNFR2, OX40, and 4-1BB [PDB IDs: 7KP8 (17), 3ALQ (17, 18), 2HEV (19), 6BWV (20)]. **(C)** (Top) Original design models and partial diffusion-generated model for the highest-affinity TNFR2, OX40, and 4-1BB binders (see also fig. S6). The two TNFR1 binders (orange, left) were superimposed on the new targets (right) and partially diffused to yield target-matched backbones (blue). (Bottom) SPR measurements show that the binders are highly specific for the targets they were diffused against.

**(D)** Crystal structure (colors) of TNFR2\_mb1 (blue) in complex with TNFR2 (yellow) superimposed on the computationally design model (gray). Boxed interface regions on the backbone superposition on the left are shown with interface side-chains in the zoom-ins on the right.

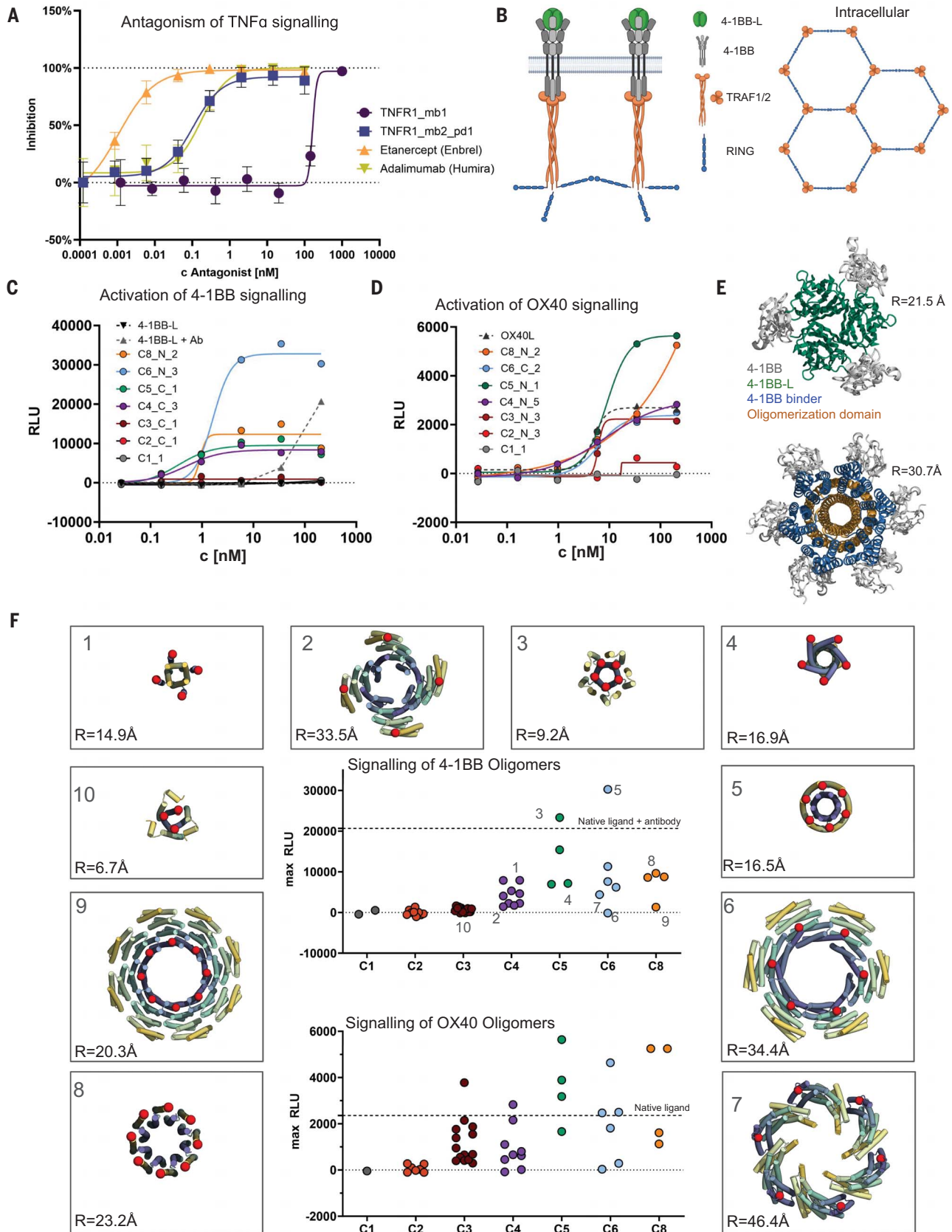


1.0 Å for designs TNFR1\_mb1 and TNFR1\_mb2, respectively, using AlphaFold2-multimer model 1], suggest that both TNFR1\_mb1 and TNFR1\_mb2 bind to TNFR1 as designed.

#### *In silico* affinity maturation

Tumor necrosis factor (TNF- $\alpha$ ) is a trimer that binds TNFR1 with high (19 pM) affinity (10), and to effectively outcompete this interaction

with a monomeric protein to counter inflammation, even higher binding affinities are required. To further optimize the affinity of the TNFR1\_mb2 and TNFR1\_mb1 designs, and



**Fig. 4. Design of soluble oligomeric 4-1BB and OX40 superagonists.** (A) Designed binders antagonize TNF- $\alpha$  signaling. HEK293-blue cells were

incubated with 100 pM TNF- $\alpha$ , and serial dilutions of designed binders and NF- $\kappa$ B-dependent activation were measured. Curves are fit to data from two

independent replicates. **(B)** Schematic overview of 4-1BB signaling. Clustering of trimeric 4-1BB ligand (4-1BB-L, green) with three copies of 4-1BB receptor (gray) leads to intracellular formation of TRAF1/2 trimers (orange) and intracellular hexamers of zinc-RING finger domains leading to downstream signaling. **(C)** Multivalent presentation of 4-1BB binder design on cyclic oligomers activates 4-1BB signaling on luciferase reporter cell lines, compared to native 4-1BB ligand (4-1BB-L) alone and complexed with an anti-his antibody. 4-1BB\_mb2 was fused to various oligomerization domains; examples for different cyclic oligomeric states with a valency of one for binder alone (C1\_1) to a valency of eight (C8\_N2) are shown. For details on oligomers, see also table S4 and fig. S10. **(D)** Multivalent presentation of OX40 binder design on cyclic oligomers activates OX40 signaling on luciferase reporter cell lines. OX40\_mb1 was fused to various oligomerization domains; examples for different cyclic oligomeric states with a valency of one for binder alone (C1\_1) to a valency of eight (C8\_N2). For details on oligomers see also

table S4 and fig. S10. **(E)** Model of 4-1BB (gray) upon binding of 4-1BB\_mb2 (blue) fused to the N terminus of a cyclic hexamer (orange) compared to native complex with 4-1BB-L (green) (right, PDB ID: 6BWV). Distance between receptor (M101) to center is indicated (R). **(F)** Geometry and oligomerization state dependence of designed 4-1BB and OX40 agonists. Binders 4-1BB\_mb2 and OX40\_mb1 were fused N- or C-terminal to 24 oligomerization domains each (see also table S4). Central panel shows maximum recorded signal at 200 pM, with each dot representing one oligomer-binder fusion, separated in groups depending on oligomeric state for C2-C8 oligomers. Each dotted line indicates signal of the native ligand (OX40) or native ligand plus antibody (4-1BB). Examples of design models of oligomerization domains for 4-1BB are shown in the surrounding panels, with numbers indicating their locations in the central plot. Fusion sites of binders are indicated by a red sphere; R is the distance of these to the center of the oligomer. Chains are colored in a gradient from N terminus (yellow) to C terminus (blue).

the TNFR1\_mb3 design, which also bound specifically (fig. S3B), rather than combining beneficial substitutions from the site saturation mutagenesis (SSM) libraries, which requires considerable experimental screening (3), we used partial diffusion (Fig. 2A): Backbones were partially noised (over 15 to 25 steps; 50 steps yield a completely random distribution), followed by RFDiffusion denoising, which yielded new backbones resembling but distinct from the original designs (RMSD 0.58 to 4.96 Å). We generated 25,000 partially diffused backbones around each starting structure, and following ProteinMPNN, the 32 designs for each starting structure that AF2 most confidently predicted bound TNFR1 in the designed binding mode (pae\_interaction <5) were selected for experimental characterization (table S1). These partially diffused designs had substantially greater CMS and buried SASA than the parent designs in all three cases (Fig. 2B). The designs were expressed in *E. coli*, and TNFR binding was measured by SPR. Most (94 out of 96) of the designs were expressed at high levels, and 30% (28 out of the 94 expressed designs) bound TNFR1 (fig. S6A). Partial diffusion increased the binding affinity of TNFR1\_mb2 by three orders of magnitude to <10 pM (Fig. 2C and fig. S6B), whereas TNFR1\_mb3 affinity increased from weak binding in the micromolar range to 20 nM (Fig. 2C). Improvements were smaller for the more regular TNFR1\_mb1 backbone; free diffusion sampling appears to have already found a close-to-ideal solution for this binding mode. For TNFR1\_mb2, the considerable increase in affinity brought about by partial diffusion likely is due to an additional interface and an overall better fit to the binding cleft with several additional contacts (Fig. 2C). The low picomolar affinity of the partially diffused TNFR1\_mb2, which we refer to as TNFR1\_mb2\_pd1 below, is considerably higher than that of any previously described monomeric TNFR1-binding protein.

### Switching specificity

Given the success of partial diffusion in increasing binding affinities, we investigated

whether a similar approach could be used to switch specificity to other TNFR family members, which are diverse in sequence but have a very similar overall fold (Fig. 3, A and B). We placed binders TNFR1\_mb2, TNFR1\_mb1, and TNFR1\_mb3 on TNFR superfamily members TNFR2, OX40, and 4-1BB (by superimposing the latter receptors on the TNFR1 in the design models) and carried out 25,000 design trajectories for each combination consisting of the addition of random Gaussian noise, RFDiffusion denoising, and ProteinMPNN sequence design. For TNFR2, 1323 designs had AF2 complex predictions with pae\_interaction <7.5—a substantially higher success rate than achieved by free diffusion on TNFR1. For each of the receptors, 48 designs were experimentally characterized (table S1). For TNFR2, 32% of the designs bound with high specificity (fig. S7A); the highest-affinity design had a  $K_D$  of 198 pM to TNFR2 and no affinity for the other family members tested (Fig. 3 and fig. S7B). Unlike TNFR2, which shares a common ligand with TNFR1 (TNF- $\alpha$ ), OX40 has a different ligand and thus a more distinct binding interface. Despite this difference in natural ligand, partial diffusion starting from the TNFR1 binders yielded an OX40 binder with a  $K_D$  of 30 nM. As expected, this binder was shifted more substantially both in approach angle and tertiary structure compared to the parental design than the TNFR2 binders (fig. S7C). For the even less related 4-1BB, an additional round of partial diffusion was required to achieve a pae\_interaction less than 7, but the experimental success rate was still high, with 22 out of 48 tested designs binding specifically to their target, with the highest affinity of 44 nM. Design 4-1BB\_mb1 illustrates how partial diffusion can conform backbones to a target, in this case, by introducing an unusual kinked helix and a short  $\beta$  sheet to pair with the distinct receptor fold (fig. S7F).

### TNFR2 design has near-atomic-level accuracy

We were able to solve the structure of the highest-affinity retargeted binder, TNFR2\_mb1, in complex with TNFR2 using x-ray crystallography (table S3). The crystal structure of the

complex is very close to that of the computational design model and nearly identical over the designed binder (C-alpha RMSD 0.52 Å; Fig. 3D, left panel). The key side-chain interactions in the extensive designed protein-protein interface are very similarly positioned as well (Fig. 3D, right panels).

### Designed binders antagonize signaling

The picomolar affinity of the TNFR1 binders makes them possible candidates for blocking inflammation. To date, targeting of the TNF- $\alpha$  pathway has primarily focused on binding to circulating TNF- $\alpha$  because antibodies targeting TNFR1, owing to their bivalency, can activate, rather than suppress, TNF- $\alpha$  signaling. We investigated whether our (monomeric) designs could inhibit TNF- $\alpha$  signaling on a TNF- $\alpha$  human embryonic kidney 293 (HEK293) reporter cell line (InvivoGen), which monitors TNF- $\alpha$  signaling via activator protein 1/nuclear factor  $\kappa$ B (AP-1/NF- $\kappa$ B)-dependent activation of a secreted embryonic alkaline phosphatase (SEAP) reporter gene (Fig. 4A). We found that the designs potentially inhibited TNF- $\alpha$  signaling, with a median inhibitory concentration (IC<sub>50</sub>) for the best design at 106 pM.

Further biophysical characterization showed that the designs have desirable developability features. They retain function after heating to 95°C for 30 min or incubation in mouse serum for 2 hours (figs. S8A and S9). Although the TNFR1 binders have substantial hydrophobic patches in their target binding surfaces (Fig. 1F), hydrophobic interaction chromatography did not show substantial hydrophobic interaction compared to Adalimumab, Etanercept, and other clinical antibodies (fig. S8B). In addition to having high specificity within the family (Fig. 3C), the highest-affinity TNFR1\_mb2\_pd1 also showed no substantial off-target binding to a TNFR1 knockout cell line at a concentration of 100 nM, 10,000 fold higher than the  $K_D$  (fig. S8C).

### Agonist design

For OX40 and 4-1BB, which have been widely studied for expanding T cells for cancer treatment, agonists could have therapeutic potential.

Unlike TNFR1, 4-IBB signaling is not activated by soluble trimeric ligands alone—the physiological ligands are in the plasma membrane of adjacent cells and induce 4-IBB arrangements with longer-range order (Fig. 4B). Signaling has been achieved by using antibody-generated ligand networks to drive higher-order complexes (11–13); although such ligands are soluble, they are also quite heterogeneous. To explore the generation of well-defined, monodisperse 4-IBB agonists, we fused the 4-IBB binder to designed homo-oligomers with different valencies and spacings between fusion sites (red dots in Fig. 4F). We found that monomers and C2 or C3 oligomers did not signal, consistent with the lack of signaling of native trimeric ligands. By contrast, signaling was observed with C4, C5, C6, and C8 oligomers (Fig. 4C and fig. S10A). The strongest signal was observed for a C6 construct, which arranges the receptors at similar distances as the native ligand, but with higher valency (six copies instead of three; Fig. 4E). Over a set of 44 oligomers (Fig. 4F), valency was the strongest determinant of signal strength, with no signal for C1–C3, a consistent but low signal for C4, and a higher signal for C5 and C6. Beyond valency, the geometry of association also influenced the extent of signaling; higher-order oligomers that separate the receptors by distances substantially greater than the native ligand exhibited low or no signal, as did those that would likely clash with the receptor or penetrate the membrane (Fig. 4F). Overall, the requirement for higher-order valencies is consistent with the proposed signaling mechanism inferred from the intracellular hexameric arrangement (14) (Fig. 4B), but it remains to be determined how bringing in just one more subunit (in the C4 case) leads to agonism.

For OX40, we again tested a range of different oligomeric states and observed a quite different pattern. In contrast to 4-IBB, trimeric constructs were effective agonists, consistent with the fact that OX40 can be activated by soluble trimeric ligands. Monomeric and dimeric binder constructs did not signal, whereas oligomeric constructs with three, four, or five binding modules efficiently activated signaling (Fig. 4, D and F, and fig. S10B).

For both 4-IBB and OX40, both the median effective concentration ( $EC_{50}$ ) and  $E_{max}$  (maximum signal) varied considerably among the oligomeric constructs, indicating a substantial opportunity for fine-tuning the response by modulating valency and geometry, both to investigate the mechanism of signaling through this important class of receptors and for therapeutic applications. Particularly interesting are the substantially higher  $E_{max}$  of the best OX40 and 4-IBB synthetic agonists compared to the native ligand in the OX40 case and antibody-ligand assemblies in the 4-IBB case; these could be particularly useful for expanding T cell populations.

## Discussion

For therapeutic challenges for which antagonism without any risk of agonism is necessary, high-affinity monomeric binders could have advantages over bivalent antibodies, which can potentially dimerize the target receptor and activate signaling (9, 15). TNF- $\alpha$  and TNFR are key drug targets given the central role this interaction plays in inflammatory disease. Current therapies primarily target the ligand TNF- $\alpha$  instead of TNFR1 to avoid potential activation of inflammatory responses, but binding TNF- $\alpha$  also inhibits potentially anti-inflammatory signaling through TNFR2, which could contribute to unwanted side effects of this important class of drugs (16). The very large interfaces of the TNFR1 binders, even larger than the native interface despite being only 107 amino acids in length, could likely enable even higher-affinity antagonism into the femtomolar range, and the high stability and likely low cost of production of small designed proteins could enable oral administration for gut disease. As illustrated by our 4-IBB and OX40 superagonists, the high-affinity monomeric binders enable the construction of a wide variety of soluble signaling molecules, offering far more control than current native ligand plus cross-linking antibody-based approaches.

More generally, the ability to generate high-affinity and high-specificity binders to therapeutically important and structurally challenging protein targets without having to immunize animals, screen large random libraries, or test thousands of design candidates ushers in a new era for binder design and therapeutic candidate discovery. The number of sequences tested (96 free diffusion and 96 partial diffusion designs for TNFR1, and 48 partial diffusion designs for TNFR2, OX40, and 4-IBB) is far fewer than in previous studies in which libraries of tens of thousands of designs were screened by using yeast display, and no random or experimentally guided optimization was involved other than selecting the best of the 96 first-round designs for partial diffusion. The <10 pM affinity for TNFR1 and 198 pM for TNFR2 are the highest that we are aware of for monomeric binders to these targets; for comparison, the antibodies that likely bind bivalently have affinities of up to 680 pM, whereas monomeric Fabs, scFvs, and nanobodies are in the range of 10 to 100 nM (15). This high affinity for targets for which multiple previous binder design efforts failed likely reflects the very high designed shape complementarity and buried surface area. Indeed, the amount of surface area these designs bury on their targets is substantially higher than in previous minibinder design efforts, and the buried surface area per residue rivals that of the native complexes, which have evolved over hundreds of millions of years (Fig. 1A). The combination of RFdiffusion starting from random residue clouds of >100 amino acids and partial

diffusion to optimize affinity and achieve high specificity for family members provides a very powerful and general approach to obtaining high-potency and high-affinity binders to challenging classes of targets.

## REFERENCES AND NOTES

1. T. Kortemme, *Cell* **187**, 526–544 (2024).
2. A. Marchand, A. K. Van Hall-Beauvais, B. E. Correia, *Curr. Opin. Struct. Biol.* **74**, 102370 (2022).
3. L. Cao *et al.*, *Nature* **605**, 551–560 (2022).
4. A. Chevalier *et al.*, *Nature* **550**, 74–79 (2017).
5. P. Gainza *et al.*, *Nature* **617**, 176–184 (2023).
6. N. F. Polizzi, W. F. DeGrado, *Science* **369**, 1227–1233 (2020).
7. J. L. Watson *et al.*, *Nature* **620**, 1089–1100 (2023).
8. S. Vázquez Torres *et al.*, *Nature* **626**, 435–442 (2024).
9. G. van Loo, M. J. M. Bertrand, *Nat. Rev. Immunol.* **23**, 289–303 (2023).
10. M. Grell, H. Wajant, G. Zimmermann, P. Scheurich, *Proc. Natl. Acad. Sci. U.S.A.* **95**, 570–575 (1998).
11. X. Yu *et al.*, *Commun. Biol.* **4**, 772 (2021).
12. C. Claus, C. Ferrara-Koller, C. Klein, *MAbs* **15**, 2167189 (2023).
13. Y. Yang *et al.*, *MAbs* **11**, 996–1011 (2019).
14. J. M. Zapata *et al.*, *Front. Immunol.* **9**, 2618 (2018).
15. R. Fischer, R. E. Kontermann, K. Pfizenmaier, *Front. Cell Dev. Biol.* **8**, 401 (2020).
16. S. Yang, J. Wang, D. D. Brand, S. G. Zheng, *Front. Immunol.* **9**, 784 (2018).
17. D. McMillan *et al.*, *Nat. Commun.* **12**, 582 (2021).
18. Y. Mukai *et al.*, *Sci. Signal.* **3**, ra83 (2010).
19. D. M. Compaan, S. G. Hymowitz, *Structure* **14**, 1321–1330 (2006).
20. R. N. Gilbreth *et al.*, *J. Biol. Chem.* **293**, 9880–9891 (2018).

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#### SUPPLEMENTARY MATERIALS

[science.org/doi/10.1126/science.adp1779](https://science.org/doi/10.1126/science.adp1779)  
Materials and Methods

Figs. S1 to S10  
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MDAR Reproducibility Checklist

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