

## BIOCHEMISTRY

## Combinatorially restricted computational design of protein-protein interfaces to produce IgG heterodimers

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Redesigning protein-protein interfaces is an important tool for developing therapeutic strategies. Interfaces can be redesigned by *in silico* screening, which allows for efficient sampling of a large protein space before experimental validation. However, computational costs limit the number of combinations that can be reasonably sampled. Here, we present combinatorial tyrosine (Y)/serine (S) selection (combYSelect), a computational approach combining *in silico* determination of the change in binding free energy ( $\Delta\Delta G$ ) of an interface with a highly restricted library composed of just two amino acids, tyrosine and serine. We used combYSelect to design two immunoglobulin G (IgG) heterodimers—combYSelect1 (L368S/D399Y-K409S/T411Y) and combYSelect2 (D399Y/K447S-K409S/T411Y)—that exhibit near-optimal heterodimerization, without affecting IgG stability or function. We solved the crystal structures of these heterodimers and found that dynamic  $\pi$ -stacking interactions and polar contacts drive preferential heterodimeric interactions. Finally, we demonstrated the utility of our combYSelect heterodimers by engineering both a bispecific antibody and a cytokine trap for two unique therapeutic applications.

## INTRODUCTION

Protein-protein interactions are important for nearly all physiological and pathological processes. One fundamental role of protein-protein interfaces is to facilitate the oligomerization of proteins, which can mediate signal transduction (1), activate or inhibit intracellular enzymes (2), and control gene expression (3). The redesign of protein-protein interfaces has emerged as an important tool for novel therapeutic development, enabling the prevention of antibiotic resistance (4) and the development of severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) (5) and G protein (guanine nucleotide-binding protein)-coupled receptor (GPCR) nanobodies (6). Advancing approaches to protein-protein interface engineering could expand therapeutic strategies for myriad diseases.

An established experimental tool for redesigning protein-protein interfaces is directed evolution. This approach relies on generating a library of variants, screening the displayed library based on a desired property, and amplifying the variants of interest (7, 8). Directed evolution mimics natural evolution, does not require previous knowledge of protein structure, and is advantageous for uncovering mutations that would otherwise be missed in rational design approaches. However, the large libraries associated with this method can be costly and cumbersome, and inefficiencies, such as low transfection rates, can prohibit the screening of the complete library

(9–13). Directed evolution is also unable to encompass all possible combinations of sequences of amino acids in an interface due to the large number of possible combinations (14). Additionally, protein-protein interfaces within obligate homomeric multimers are generally recalcitrant to experimental directed evolution methods since the protein subunits cannot exist or are too destabilized to display and capture in isolation.

Because of these challenges, *in silico* redesign of protein-protein interfaces represents a promising alternative for certain protein-protein interfaces. In such computational approaches, neural networks generated by machine learning such as AlphaFold (15, 16) and RoseTTAFold (17, 18) can be combined with *in silico* screening methods to efficiently screen and select a large number of variants based on a specific property before experimental validation. Such a strategy has been used to identify protein-protein interaction hotspots using the interface mode of the Rosetta software package, wherein potentially interacting residues within a protein-protein interface were identified based on distance, mutated *in silico* to alanine, and the calculated change in binding free energy ( $\Delta\Delta G$ ) was used as a predictor for the change in binding (19).

Expanding this strategy to sample all 20 amino acids at each residue, while technically possible, is not feasible due to the amount of time and computational power required. For instance, conducting a combinatorial computational search of all 20 amino acids at each of just 10 positions within a protein-protein interface would require some 10 billion  $\Delta\Delta G$  calculations, which at 1 s per calculation would require more than 2500 years of computational time using 128 cores. While future advances in computational speed and parallelization will eventually make such *in silico* screening approaches possible, it is currently necessary to limit computational expense and, therefore, the *in silico* library of variants tested.

One way in which to reduce the chemical space to be screened, and the resulting computational cost, is to apply previous knowledge of the specific protein-protein interface, or of protein-protein interfaces in general. It has been previously reported that the

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complementarity determining regions (CDRs) of antibodies are enriched in tyrosines and serines (20–22) and that antibodies with high affinity against the human death receptor (DR5), the human vascular endothelial growth factor (hVEGF), and others could be engineered using a phage display library consisting of only tyrosine and serine residues in their CDR regions (23–28). Other proteins have also been engineered in a similar manner, with the fibronectin type III domain (FN3) being engineered to bind maltose-binding protein (MBP), human small ubiquitin-like modifier 4 (hSUMO4), and yeast small ubiquitin-like modifier ( $\gamma$ SUMO) (29, 30).

Inspired by these experimental successes using highly restricted variant libraries, we hypothesized that combining the Rosetta  $\Delta\Delta G$  approach described above with an *in silico* library restricted to tyrosine and serine mutations would allow us to efficiently, and sufficiently, sample the effects of varying multiple positions within a protein-protein interface to identify combinations of mutations that either improve or worsen binding. Specifically, we sought to remodel the binding interface between the fragment crystallizable (Fc) protomers of an immunoglobulin G1 (IgG1) antibody to promote the favorable formation of heterodimers while making any potential resulting homodimers energetically unfavorable. Because engineered IgG heterodimers can exhibit improved therapeutic efficacy and expand the possibilities for constructing Fc-fusion proteins, methods to redesign the IgG Fc homodimer interface to produce Fc heterodimers with unique mutations in either Fc protomer have been previously established, including but not limited to “knobs-in-holes” (KiH), which relies on hydrophobic and steric interactions (31); electrostatic steering, which employs charged side chains (32); and a multistate design strategy using negative state repertoires to screen for mutations that promoted heterodimer formation while disfavoring homodimer formation (33).

Here, we combined restricted libraries with *in silico* screening to develop a computational approach to redesigning protein-protein interfaces called combYSelect. Using our combYSelect approach, we designed two IgG1 heterodimers that exhibit similar heterodimeric chain pairing and stability compared to previously reported IgG Fc heterodimers. We established the molecular basis of heterodimerization for both of these heterodimers and presented two functional applications of these heterodimeric IgG1 antibodies.

## RESULTS

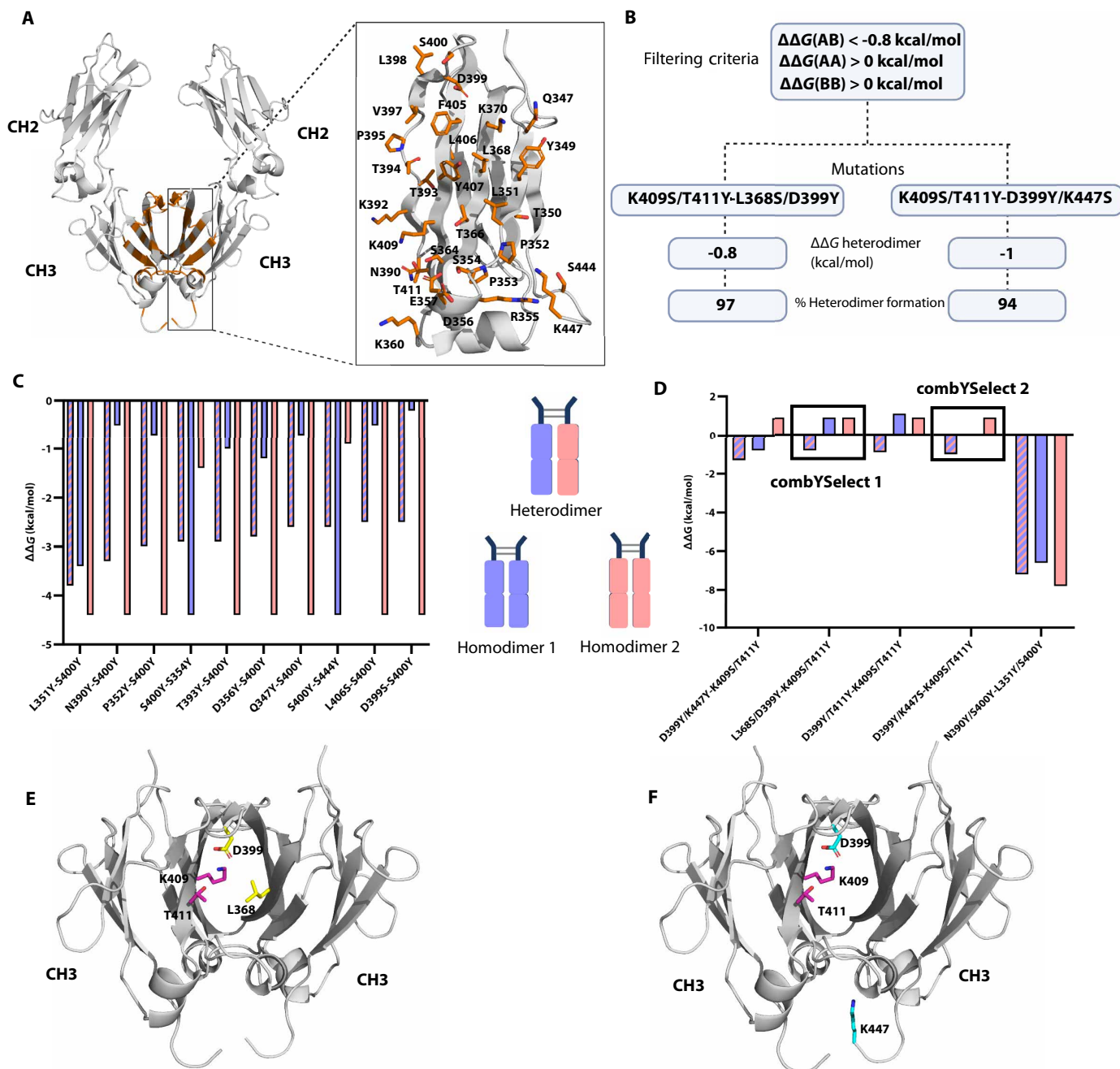
### Serine and tyrosine mutations at numerous positions promote Fc heterodimer formation

To identify mutations in the CH3 domain of the IgG1 heavy chain (Hc) that would promote heterodimer formation while preventing homodimer formation, we established a computational strategy that we named combYSelect. In this strategy, we identified all residues in the CH3 domain of an IgG1 Fc within 4 Å, as a cutoff for interacting residues, of the opposing CH3 domain. Since the last three to four amino acids of the C terminus are not resolved in any published IgG1 Fc structures, we used an AlphaFold model of an IgG1 Fc to screen all possible residues. We focused exclusively on residues within the CH3 region since these were most likely to be functionally inert compared to residues in the upper regions of the CH2 domains and the hinge, which engage in Fc $\gamma$  receptor (Fc $\gamma$ R) binding as part of the immune cascade (34). We identified a total of 31 residues that met these criteria: Q347, Y349, T350, L351, P352, P353,

S354, R355, D356, E357, K360, S364, T366, L368, K370, N390, K392, T393, T394, P395, V397, L398, D399, S400, F405, L406, Y407, K409, T411, S444, and K447 (Fig. 1A). We used a combinatorially restricted library, testing mutations only to serine and tyrosine. We chose these two amino acids based on previous studies that developed high-affinity fragment antigen-binding (Fabs) and scaffold proteins with phage display libraries of only serine and tyrosine residues (23, 26). This allowed us to increase the number of mutations we could assess while maintaining reasonable computational times. We carried out the screen in a combinatorial manner by simultaneously mutating one or two residues in each chain. Mutations of one residue per chain yielded 3844 mutations and took 5 min using 64 central processing unit (CPU) cores. Mutations of two residues per chain yielded 3,459,600 mutations and took 15 hours on 64 CPU cores. We did not test three mutations per chain since, even with the restricted serine/tyrosine library, such a calculation would have taken over 8 months using 64 CPU cores.

To computationally assess the effect of our mutations on heterodimerization, we calculated the  $\Delta\Delta G$  for each combination of mutations using the interface mode of Rosetta (19), which outputs a change in  $\Delta G$ , or  $\Delta\Delta G$  value, where positive values indicate that the mutation reduces stability of the protein complex, while negative values signify improved binding relative to the wild-type (WT) model. We chose  $\Delta\Delta G < -0.8$  kcal/mol as the criterion for a potential successful mutation to account for the previously observed variability between experimentally measured and computationally predicted  $\Delta\Delta G$  values from that protocol (19, 35). Additionally, while developing and optimizing this strategy, we determined that a predicted favorable value for  $\Delta\Delta G_{\text{heterodimer}}$  was not sufficient as a selection criterion. As a result, we added a filtering step in which we selected for  $\Delta\Delta G_{\text{homodimer}} > 0$  to ensure that we selected mutants in which homodimer formation was energetically unfavorable (Fig. 1B).

When mutating one chain per residue, we observed that the 10 mutants with the most negative  $\Delta\Delta G_{\text{heterodimer}}$  values also had negative  $\Delta\Delta G_{\text{homodimer}}$  values (Fig. 1C). Thus, these mutants did not fit our criteria and will likely still form an appreciable proportion of homodimers. We therefore assessed the results of the *in silico* mutations of two residues per chain. We selected a total of five sets of mutations from the screen that resulted in negative  $\Delta\Delta G_{\text{heterodimer}}$  values (Fig. 1D). The first set of mutations D399Y/K447Y-K409S/T411Y had a  $\Delta\Delta G_{\text{heterodimer}}$  of  $-1.3$  kcal/mol, and one of the homodimers was energetically unfavorable with a  $\Delta\Delta G_{\text{homodimer } 2}$  of 0.9 kcal/mol; however,  $\Delta\Delta G_{\text{homodimer } 1}$  was  $-0.8$  kcal/mol. Another set of mutations N390Y/S400Y-L351Y/S400Y had the most negative  $\Delta\Delta G_{\text{heterodimer}}$  of  $-7.2$  kcal/mol. However, the similarly negative  $\Delta\Delta G_{\text{homodimer}}$  values made it an unsuitable candidate. Only three of the variants followed all the filtering criteria of negative  $\Delta\Delta G_{\text{heterodimer}}$  and positive  $\Delta\Delta G_{\text{homodimer}}$  values: L368S/D399Y-K409S/T411Y ( $\Delta\Delta G_{\text{heterodimer}} = -0.8$  kcal/mol,  $\Delta\Delta G_{\text{homodimer } 1} = 0.9$  kcal/mol,  $\Delta\Delta G_{\text{homodimer } 2} = 0.9$  kcal/mol), D399Y/T411Y-K409S/T411Y ( $\Delta\Delta G_{\text{heterodimer}} = -0.9$  kcal/mol,  $\Delta\Delta G_{\text{homodimer } 1} = 1.1$  kcal/mol,  $\Delta\Delta G_{\text{homodimer } 2} = 0.9$  kcal/mol), and D399Y/K447S-K409S/T411Y ( $\Delta\Delta G_{\text{heterodimer}} = -1$  kcal/mol,  $\Delta\Delta G_{\text{homodimer } 1} = 0$  kcal/mol,  $\Delta\Delta G_{\text{homodimer } 2} = 0.9$  kcal/mol). We selected for further experimental testing the two candidates that each had four unique mutations: L368S/D399Y-K409S/T411Y—hereafter referred to as “combYSelect1” (Fig. 1E) and D399Y/K447S-K409S/T411Y—hereafter referred to as “combYSelect2” (Fig. 1F).



**Fig. 1. combYSelect screening strategies and workflow.** (A) AlphaFold model of IgG1 Fc that was used for in silico mutation screen, with residues identified to be 4 Å apart from another residue on the opposing chain highlighted in orange stick form. (B) Schematic depicting the combYSelect strategy used to screen for mutations that promote IgG1 heterodimer formation. We used the interface mode of Rosetta coupled with restricting mutations to tyrosine (Y) and serine (S) and filtering based on  $\Delta\Delta G_{\text{homodimer}}$  and  $\Delta\Delta G_{\text{heterodimer}}$  values. (C) Bar graph depicting the computationally determined  $\Delta\Delta G$  values for heterodimers (blue-pink stripes) or homodimers (blue, when the mutation listed first is considered, and pink when the mutation listed second is considered) of Fcs, in which only one residue on each chain was mutated. (D) Bar graph depicting the computationally determined  $\Delta\Delta G$  values for heterodimers (blue-pink stripes) or homodimers (blue, when the set of mutations listed first is considered, and pink when the set of mutations listed second is considered) of Fcs, in which two residues on each chain were mutated. (E) The two sets of mutations for combYSelect1: K409S/T411Y-L368S/D399Y, one of the heterodimers selected for further testing, are highlighted in purple and yellow, respectively. (F) The two sets of mutations for combYSelect2: K409S/T411Y-D399Y/K447S, one of the heterodimers selected for further testing, are highlighted in purple and cyan, respectively.

### Mutations identified by combYSelect result in maximal heterodimerization

To experimentally assess the relative yields of heterodimer and homodimer formation for our combYSelect designs and compare them to published KiH (T366W-T366S/L368A/Y407V) (36) and electrostatic steering (E357Q/S364K-L368D/K370S) heterodimers (32), we developed a model system involving the coexpression of an IgG1 Fc region and full-length Hc of rituximab, each from distinct protomers in a heterodimer design, as well as the light chain (Lc) of rituximab, our model IgG1 antibody. Rituximab, an anti-CD20 chimeric monoclonal antibody, targets B cells and is often used as a therapeutic for B cell lymphomas (37). After purification by protein A affinity chromatography (38) and treatment with the IgG-specific endoglycosidase EndoS2 (39, 40) to remove heterogeneous glycosylation at residue Asn297, three potential products with unique masses that are easily detectable by liquid chromatography–mass spectrometry (LC-MS) can result from such a coexpression, including an intact Fc (~50 kDa), an intact IgG (~150 kDa), and a monovalent IgG molecule (Fab<sub>1</sub>Fc; ~100 kDa) (Fig. 2A). The formation of intact Fc regions and IgG antibodies each result from the two possible homodimeric pairs, while the formation of the Fab<sub>1</sub>Fc molecule arises strictly from heterodimerization. Thus, the ratio of Fab<sub>1</sub>Fc to Fc and IgG is a direct measure of heterodimer formation.

When we evaluated WT rituximab in this assay, we observed 63% Fab<sub>1</sub>Fc (Fig. 2B). For the KiH heterodimer (T366W-T366S/L368A/Y407V) (36), we observed 99% heterodimerization when the Hc, containing the T366S/L368A/Y407V mutations, was coexpressed with the Fc containing the T366W mutation, along with the Lc (Fig. 2C, left). Similarly, we observed 99% Fab<sub>1</sub>Fc formation when the T366W mutation was on the Hc and the T366S/L368A/Y407V mutations were on the Fc (Fig. 2C, right). The electrostatic steering heterodimer we tested resulted in 91% Fab<sub>1</sub>Fc formation regardless of which chain contained each set of mutations (Fig. 2D). In comparison to the controls above, we observed 97% Fab<sub>1</sub>Fc formation for both mutation-chain combinations for combYSelect1 (L368S/D399Y-K409S/T411Y) (Fig. 2E). Similarly, combYSelect2 (D399Y/K447S-K409S/T411Y) resulted in 96% Fab<sub>1</sub>Fc formation when the D399Y/K447S mutations were contained within the Hc (Fig. 2F, left) and 91% when that set of mutations was contained in the Fc (Fig. 2F, right). These results indicate that our combYSelect heterodimers exhibit nearly ideal IgG heterodimerization that is comparable to those of two other previously identified heterodimers that we selected for testing.

### Heterodimer mutations do not affect IgG stability or FcRn binding

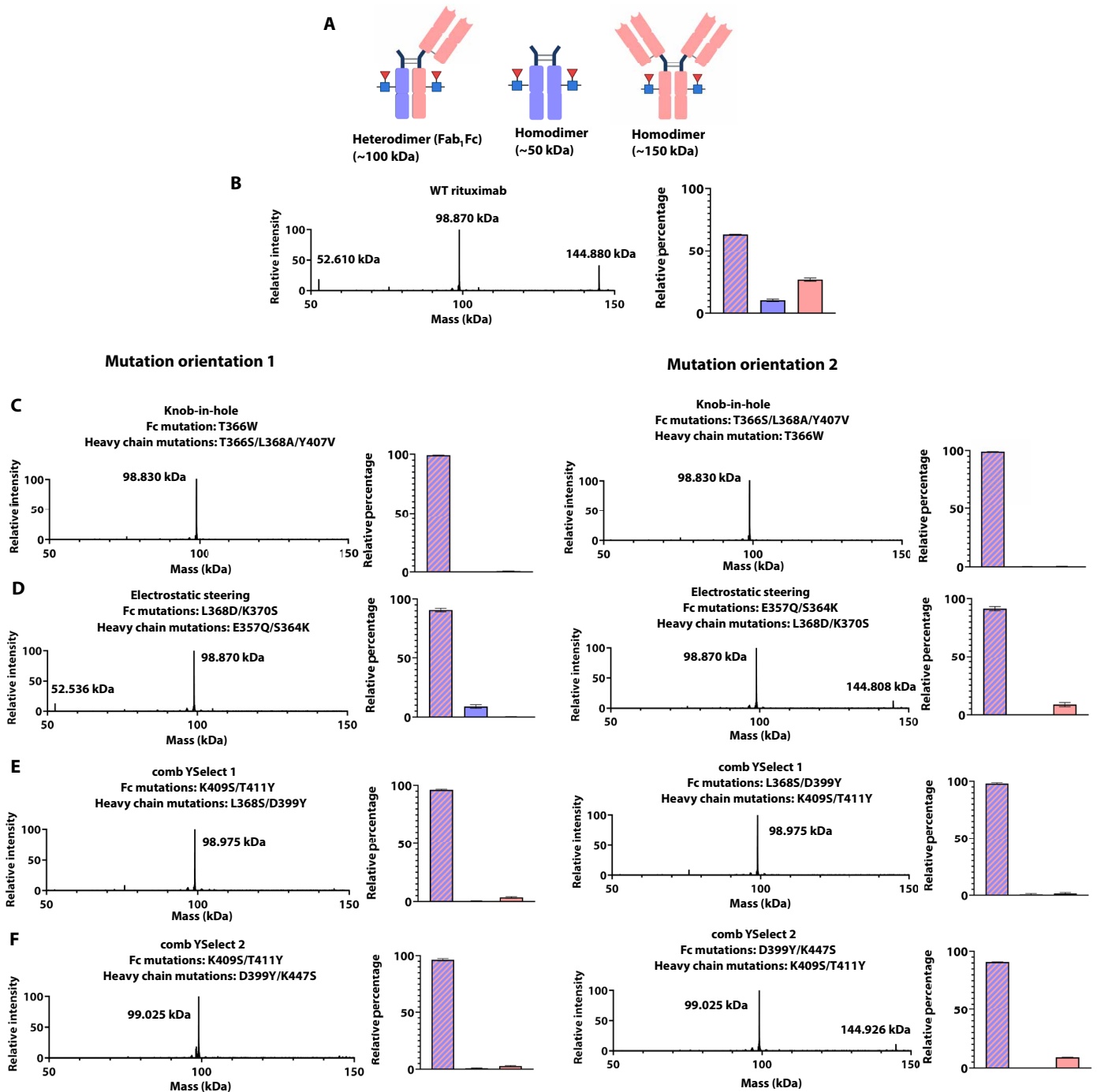
In addition to promoting heterodimer formation, we wanted to ensure that the mutations introduced in the CH3 domains did not adversely affect the functional properties of IgG molecules. We assessed the thermal stability of each construct by determining the melting curves of the resulting heterodimeric antibodies in the Fab<sub>1</sub>Fc form (i.e., the product resulting from coexpression of the Hc, Lc, and Fc fragment) using nano-differential scanning fluorimetry (nDSF), which, unlike conventional DSF, is a dye-free method that relies on the detection of tryptophan autofluorescence as a function of temperature (41). Our results showed that rituximab had three transition temperatures: 74.9°C, 81.8°C, and 90°C, corresponding to the unfolding of the CH2, Fab, and CH3 domains, respectively (Fig. 3, A and B, and fig. S1A), which is in agreement with

previously reported results (42). Additionally, the near-identical transition temperatures between the rituximab IgG and rituximab in the Fab<sub>1</sub>Fc form confirmed that thermal stability of the latter is representative of that of the full-length IgG. In contrast, the KiH (36) (Fig. 3C) and electrostatic steering constructs (Fig. 3D) previously reported in the literature (32) and combYSelect1 and combYSelect2 (Fig. 3, E and F), all of which are in the Fab<sub>1</sub>Fc form, only showed two transition temperatures corresponding to the CH2 and Fab domains. The first transition temperature ranged from 74.1°C to 76.3°C, whereas the second was between 80.8°C and 83.1°C. When we assessed the melting temperatures of combYSelect1 and combYSelect2 Fcs, we observed only one melting temperature at approximately 74°C, which usually corresponds to the CH2 domain (fig. S1, B and C). This suggests that the CH3 domain of the heterodimers we tested either unfolds at the same temperature as the CH2 domain or only undergoes reversible unfolding within our tested temperature range.

We also assessed the potential effects of these mutations on the overall structure of the CH2-CH3 interface. This region is of particular interest as it is where the neonatal Fc receptor (FcRn) binds in a pH-dependent manner to recycle endogenous IgGs and extend their serum half-life (43). We used biolayer interferometry (BLI) analysis to compare the FcRn binding affinity of rituximab to combYSelect1 and combYSelect2 IgGs with rituximab Fab domains. We obtained nearly identical dissociation constant ( $K_D$ ) values of 14.3, 17.0, and 8.0 nM for rituximab, combYSelect1, and combYSelect2, respectively, at pH 6 (Fig. 3G and fig. S2, A to F). As expected, all of these IgGs showed minimal binding to FcRn at pH 7.4 (fig. S2, G to I). Overall, our data show that the heterodimer forming mutations did not affect the stability or function of the IgG.

### Conformationally variant residues exist in the heterodimer interface

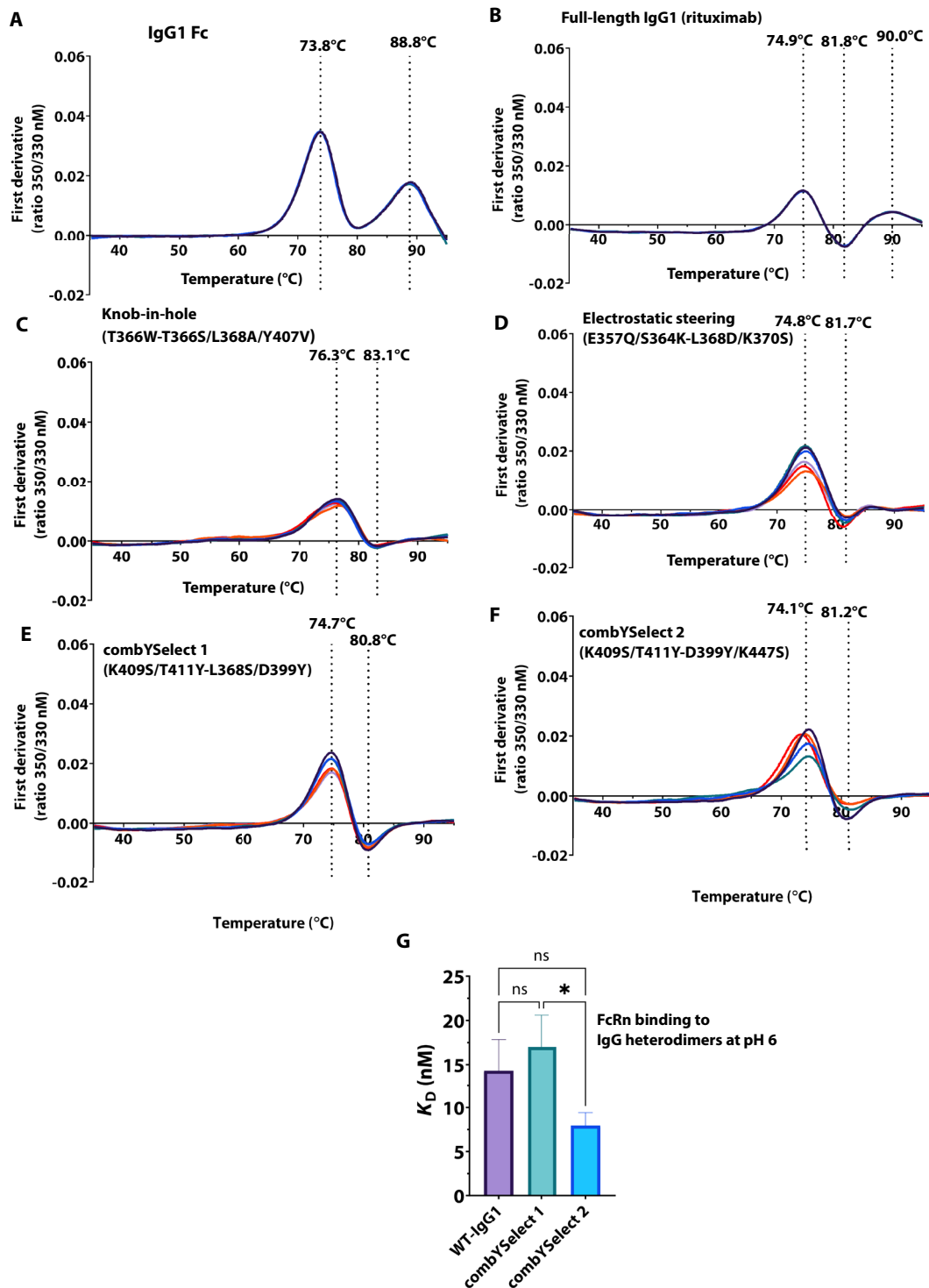
To understand how the introduced mutations promote heterodimer formation, we determined the x-ray crystal structures of combYSelect1 and combYSelect2 Fc regions to resolutions of 2.5 Å and 3 Å, respectively (Fig. 4 and table S1). These structures revealed that the T411Y mutation in the first protomer is involved in  $\pi$ -stacking interactions with the D399Y mutation on the opposing protomer (Fig. 4, A and B, left panels). The electron density map shows continuous unaccounted electron density near the mutated tyrosines, which was best explained by modeling these residues as occupying multiple conformations and positions. We modeled two possible conformations for each mutated tyrosine in combYSelect1 (fig. S3A). In conformation A, the hydroxyl groups of T411Y and D399Y are 2.3 Å apart and likely form a hydrogen bond. Similarly, when D399Y is in conformation B, its hydroxyl group is at a distance from the carboxyl group of N390 on the opposing chain, consistent with the formation of polar contacts with the amide group (Fig. 4A). We observed similar interactions in combYSelect2, in which unmodeled electron density near mutation D399Y also suggested that this residue likely occupies numerous positions. We similarly modeled two possible conformations. However, we observed no such density for the T411Y mutation and therefore modeled only one tyrosine rotamer conformation in the interface. (Fig. 4B and fig. S3B). It is important to note that our data do not enable us to identify specific conformations or pairings of the tyrosine residues. Instead, the modeled conformations represent some of the states that can exist transiently. Still, such alternate conformations of interface residues



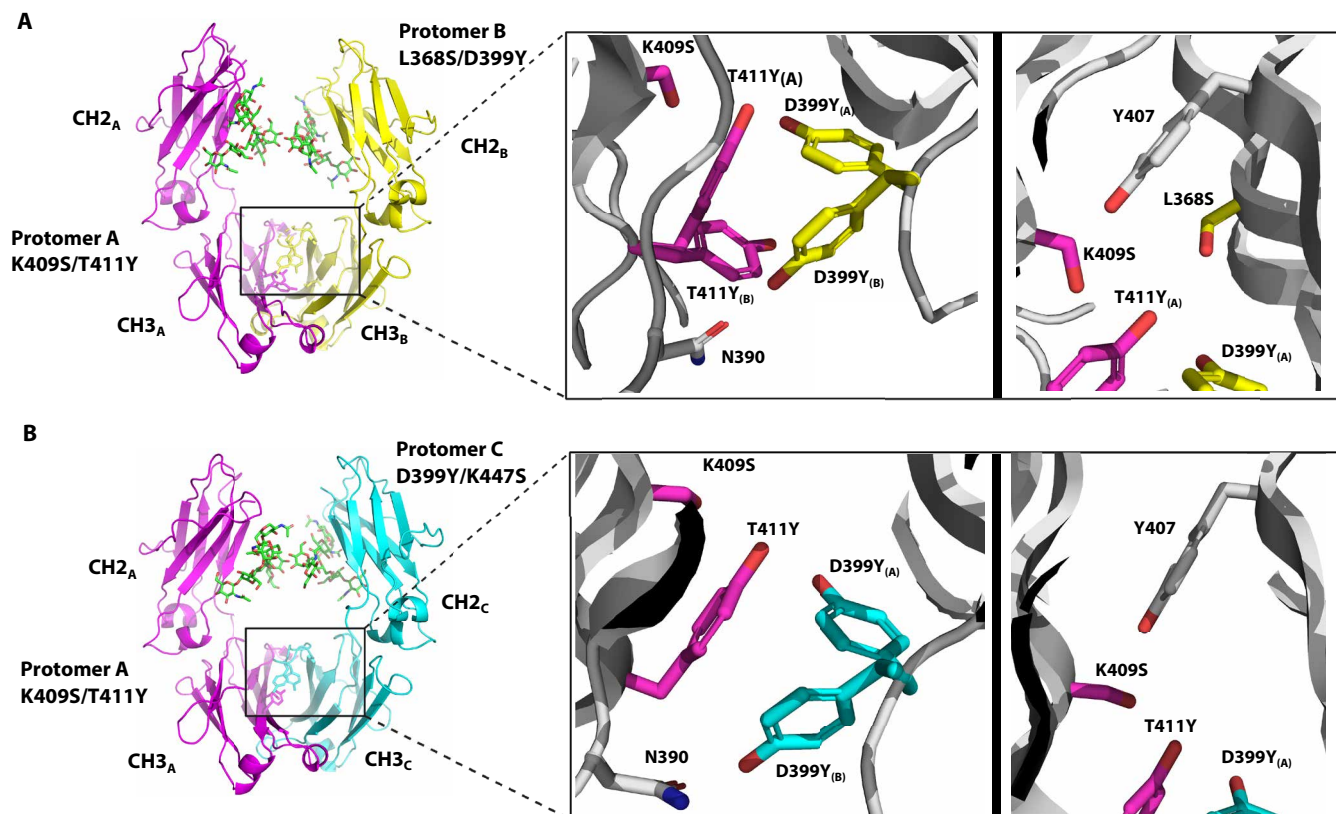
**Fig. 2. Intact LC-MS heterodimer assessment.** (A) Diagram showing the potential constructs formed upon coexpression of the Hc, Lc, and Fc plasmids. MS spectra depicting homodimer (Fc or IgG) and heterodimer formation of Fab<sub>1</sub>Fc for WT rituximab (B), KiH (C), electrostatic steering (D), combYSelect1 (E), and combYSelect2 (F). The intensity of each of the three peaks is determined relative to each other, with the highest peak normalized to 100. The bar graphs depict the relative percentage formation of the Fab<sub>1</sub>Fc heterodimer (blue with pink stripes), Fc homodimer (blue), and IgG homodimer (pink).

are consistent with other published structures of IgG heterodimers. The EW-RVT heterodimer was improved by introducing a disulfide bond in the CH3 domain (44), perhaps counteracting the dynamic tendencies of the original heterodimer. The structure of the ZW1 heterodimer, which is formed by four distinct mutations in each chain, was modeled with each of the mutated residues at a 50%

occupancy for each of two possible CH3/CH3 heterodimeric orientations (45). This suggests that the Fc chains can consistently fluctuate between multiple possible orientations. Additionally, the hydroxyl moieties of the mutated S409 residue on protomer A and Y407 on protomers B and C are 3.6 Å and 2.8 Å apart, respectively, and form potential hydrogen bonds (Fig. 4, A and B, right panels).



**Fig. 3. Stability and function of IgG heterodimers.** (A to F) First derivative plots of the melting temperatures of the Fab<sub>1</sub>Fc forms of each of the indicated constructs. The labeled temperatures are the mean of six total replicates, three for each orientation in which the mutations per chain are on the Fc fragment only, or on the Fc and Fab fragment. (G)  $K_D$  values for WT IgG1, combYSelect1, and combYSelect2 IgG derived from BLI binding assays. Statistical significance was determined by one-way ANOVA with Tukey's multiple-comparison test (ns  $P > 0.05$ , \* $P \leq 0.05$ , \*\* $P \leq 0.01$ , \*\*\* $P \leq 0.001$ , \*\*\*\* $P \leq 0.0001$ ).



**Fig. 4. Structure of combYSelect1 and combYSelect2.** (A) X-ray crystal structure (2.5 Å) highlighting the T411Y/K409S and L368S/D399Y mutations. The tyrosines are dynamic with two conformations modeled for T411Y at 67% (T411Y<sub>A</sub>) and 33% (T411Y<sub>B</sub>) occupancy and for D399Y at 42% (D399Y<sub>A</sub>) and 58% (D399Y<sub>B</sub>) occupancy. (B) X-ray crystal structure (3 Å) highlighting the T411Y/K409S and D399Y/K447S mutations. The tyrosines are dynamic with two conformations modeled for D399Y at 38% (D399Y<sub>A</sub>) and 62% (D399Y<sub>B</sub>) occupancy. N390 and Y407 are nearby residues involved in the interactions.

Overall, our structures reveal that the introduced residues in our combYSelect IgGs exist in conformations that likely engage in  $\pi$ - $\pi$  interactions and H-bonds to drive heterodimer formation.

#### All combYSelect mutations are required for optimal heterodimerization

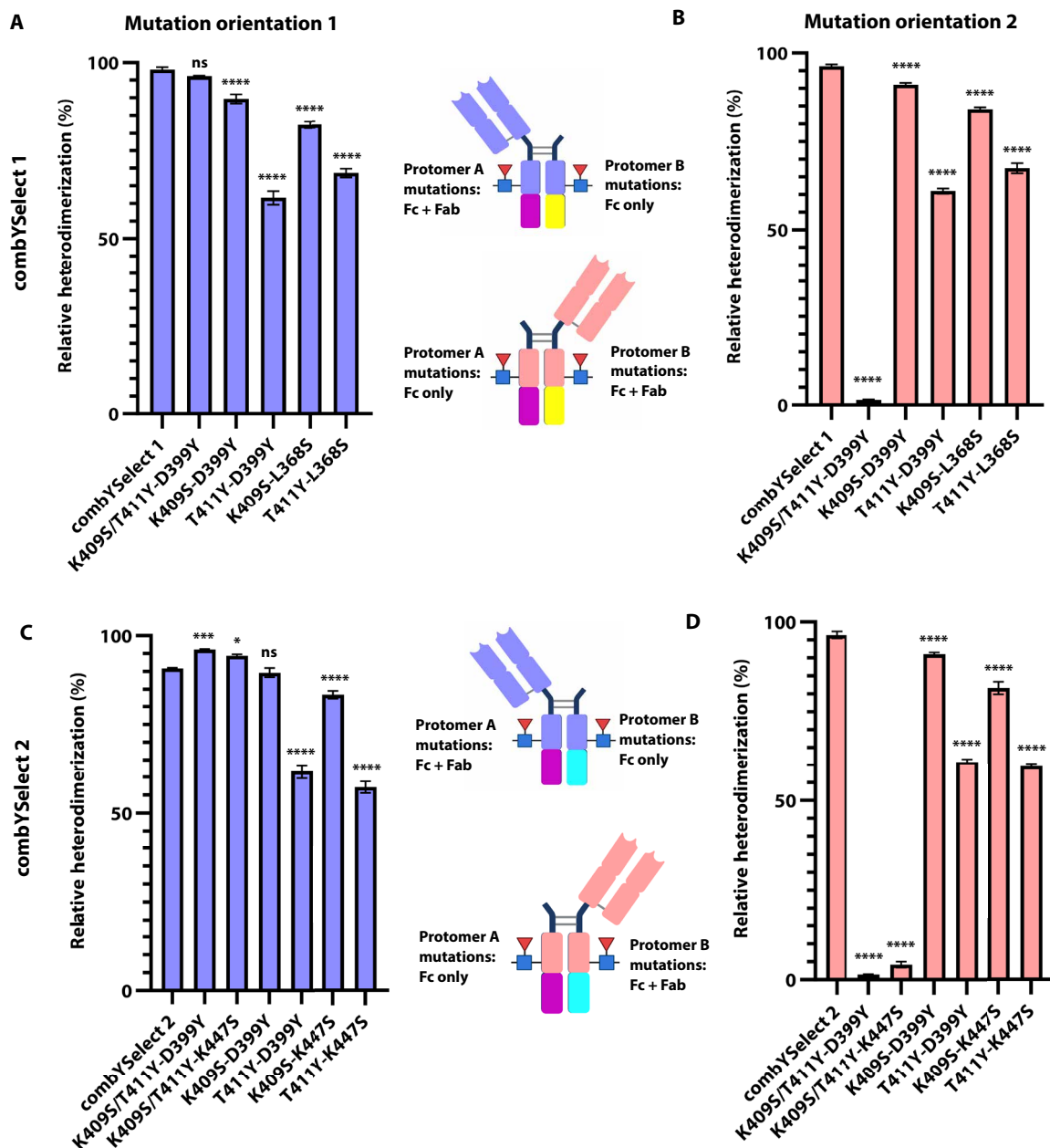
Our structure suggests that the L368S mutation in combYSelect1 is not at a sufficient distance to make any polar contacts with the residues on the opposing chain. Additionally, the electron density in our structure of combYSelect2 was insufficient to model K447S. As a result, we used the same intact LC-MS method for assessing heterodimerization described in Fig. 2 to determine whether those two mutations are essential for optimal heterodimerization. The L368S mutation was dispensable when the D399Y mutation was on an Fc, with protomer A mutations on the Hc. The observed 96% heterodimerization for K409S/T411Y-D399Y was not significantly different from combYSelect1 (Fig. 5A and fig. S4A, right). However, there was no detectable heterodimerization when the D399Y mutation was on the Fc fragment attached to the Fab, suggesting that L368S might play a role in stabilization of the Hc (Fig. 5B and fig. S4A, left). On the other hand, we observed 96% and 94% Fab<sub>1</sub>Fc formation for the K409S/T411Y-D399Y and K409S/T411Y-K447S Hc-Fc constructs, respectively, which indicated improved heterodimerization compared to combYSelect2 (Fig. 5C and fig. S4, A and B, right). However, when the D399Y and K447S individual mutations were

contained in the Hc, they were required for heterodimer formation (Fig. 5D and fig. S4, A and B, left). Finally, we confirmed that constructs with single mutations on each chain in all four possible combinations—K409S-D399Y, T411Y-D399Y, K409S-L368S, and T411Y-L368S for combYSelect1 and K409S-D399Y, T411Y-D399Y, K409S-K447S, and T411Y-K447S for combYSelect2—have significantly reduced heterodimerization regardless of mutation-chain pairing (Fig. 5 and fig. S4, C to J).

#### IgG1 heterodimers predicted by combYSelect have therapeutic potential

IgG heterodimers are often used for the development of bispecific monoclonal antibodies and Fc-fusion proteins that exhibit unique and/or improved therapeutic properties (46). To confirm that combYSelect heterodimers can be used broadly for potential therapeutic applications, we engineered combYSelect-based bispecific antibodies and cytokine traps and tested their abilities to bind simultaneously to two target cell lines expressing distinct antigens and to inhibit cytokine-mediated cellular signaling, respectively.

First, we designed a bispecific antibody in which one arm consisted of the rituximab anti-CD20 Fab, while the other was a nanobody specific for human epidermal growth factor receptor 2 (HER2) (Fig. 6A). Both of our bispecific combYSelect heterodimers were able to simultaneously bind the two antigens expressed on Raji (CD20<sup>+</sup>) and BT474 (HER2<sup>+</sup>) cells, as shown by the formation of

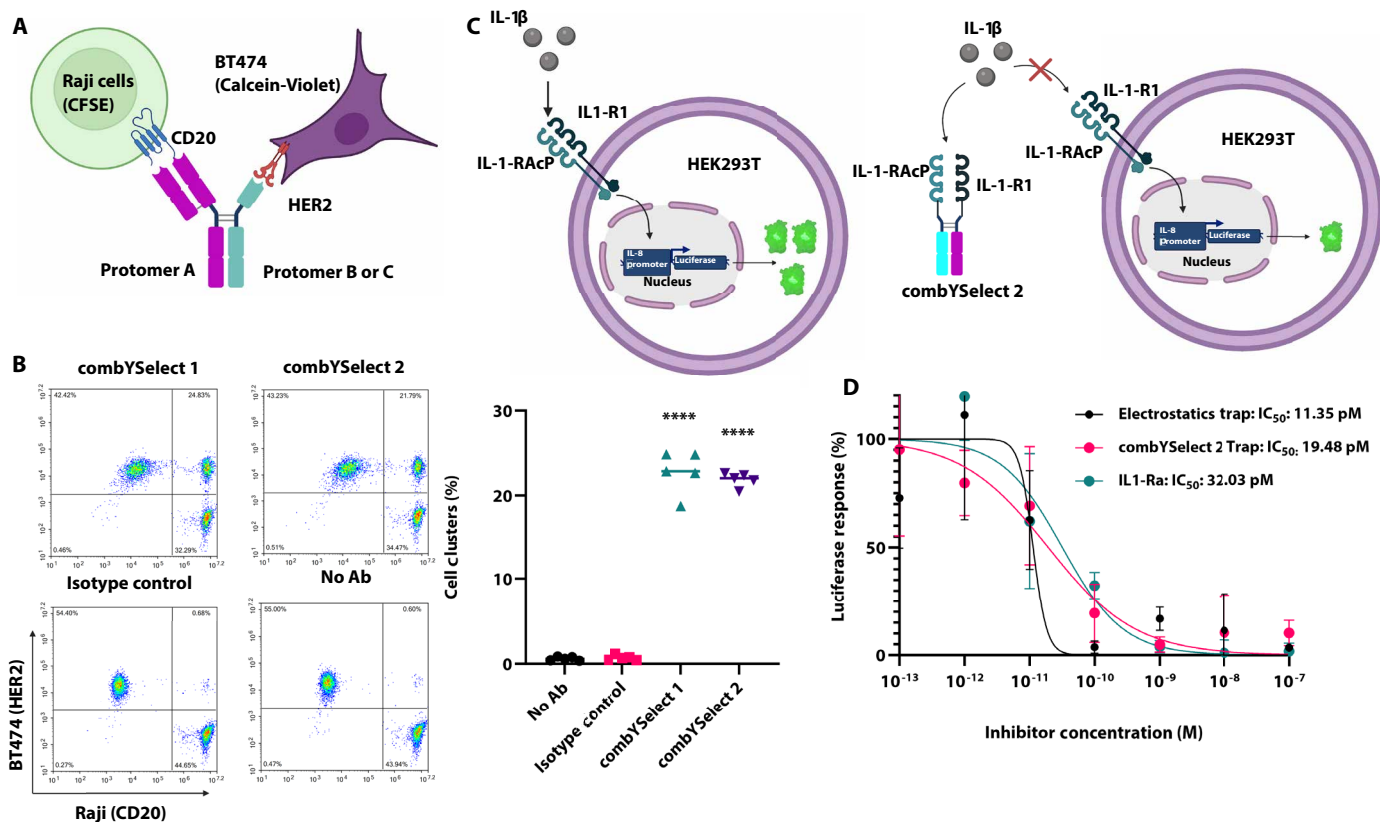


**Fig. 5. Mutation controls for combYSelect1 and combYSelect2.** (A and B) Histograms of the relative percentage of heterodimerization as determined by intact LC-MS when some or all protomer A mutations are on the fragment containing a Fab and Fc and protomer B mutations are on the fragment containing only an Fc (A) or when the orientation is reversed (B). (C and D) Histograms of the relative percentage of heterodimerization as determined by intact LC-MS when some or all protomer A mutations are on the fragment containing a Fab and Fc and protomer C mutations are on the fragment containing only an Fc (C) or when the orientation is reversed (D). Statistical significance ( $n = 3$  for each construct) was determined by one-way ANOVA with Dunnett's multiple-comparison test, in which each mutation control was compared to combYSelect1 or combYSelect2 (ns  $P > 0.05$ , \* $P \leq 0.05$ , \*\* $P \leq 0.01$ , \*\*\* $P \leq 0.001$ , \*\*\*\* $P \leq 0.0001$ ).

cell clusters that were positive for the distinct dyes used to stain each of the two cell types (Fig. 6B and fig. S5A). This was the case regardless of the antigen specificity of each protomer in versions of these bispecific antibodies in which the anti-CD20 Fab and anti-HER2 nanobody were fused to the opposite Fc protomer (fig. S5, B and E). Conversely, cell-bridging clusters did not form when using a non-specific IgG1 isotype control (Fig. 6B) or any monospecific heterodimer or homodimer (fig. S5, C to E).

IgG heterodimers are also commonly used in Fc-fusion-based therapeutics. Accordingly, in a second potential therapeutic scenario, we used the combYSelect2 Fc to construct a cytokine trap composed of the interleukin-1 (IL-1) cytokine cognate receptor IL-1 receptor I (IL-1RI) and secondary receptor IL-1 receptor accessory protein (IL-1RAcP) to sequester IL-1 $\beta$  (Fig. 6C). We measured the ability of our combYSelect2 cytokine trap to sequester IL-1 $\beta$ , and in turn inhibit luciferase expression driven by the IL-8 promoter in human





**Fig. 6. Functional applications of combYSelect heterodimers.** (A) Representation of the cell-bridging assay, in which the Fc of protomer A (combYSelect1 and combYSelect2) is engineered with an anti-CD20 rituximab Fab and the Fc protomer B (combYSelect1) or protomer C (combYSelect2) is attached to an anti-HER2 nanobody. Binding to CFSE-stained Raji cells and Calcein-Violet BT474 cells will be assessed. (B) Flow cytometry density plots and a scatterplot depicting cell cluster formations of Raji and BT474 cells when combYSelect1, combYSelect2, a nonspecific IgG1 isotype control, or no antibody was added to the cell mixture. Statistical significance ( $n = 5$ ) was determined by one-way ANOVA with Tukey's multiple-comparison test ( $ns P > 0.05$ ,  $*P \leq 0.05$ ,  $**P \leq 0.01$ ,  $***P \leq 0.001$ ,  $****P \leq 0.0001$ ). Each combYSelect IgG was compared to both the no antibody and isotype control. (C) Depiction of combYSelect2 cytokine trap and the luciferase inhibition cell-based assay. (D) Percentage of luciferase response, normalized to that when no inhibitor is added, is plotted as a function of varying inhibitor concentration in the presence of 5 pM IL-1 $\beta$ . The IC<sub>50</sub> values were determined using a nonlinear least square fit.

embryonic kidney (HEK) 293T cells, which express IL-1RI and IL-1RAcP endogenously (47). Inhibition by combYSelect2 trap was comparable to that of a trap fused to a published electrostatic heterodimer E357Q/S364K-L368D/K370S (32). Both traps had slightly improved inhibition compared to IL-1 receptor antagonist (IL-1Ra; Fig. 6D), the natural antagonist of IL-1 cytokine signaling, (48) the recombinant form of which is Anakinra that is used to treat rheumatoid arthritis and other chronic inflammatory conditions (49, 50).

## DISCUSSION

Protein-protein interfaces are large and complex, involving many individual amino acid residues that collectively exhibit heterogeneous energetic landscapes to dictate binding affinities, kinetics, and thermodynamics. The ability to accurately model and predict energetic interactions between protein interfaces *in silico* has led to great success in the protein engineering field. However, the sample space of most proteins currently remains too large to be screened comprehensively. Although increases in computing power are expected, it is unlikely that they will be able to sample the complete diversity of protein sequences in a timely manner. The recent increase in the

number of artificial intelligence-based models for protein design (51) and predicting the consequences of mutations in proteins (52) may enable us to bridge the gap between current computational capabilities and efficiently sampling the complete range of protein diversity, potentially leading to routine *de novo* protein design and engineering in the future. However, the design and redesign of protein-protein interfaces currently remains a challenge.

One strategy to overcome this challenge is to reduce the number of calculations to be made by reducing the combinations of amino acid positions within a given interface and the number of possible amino acid mutations to be made at each position. While sampling too many combinations is computationally infeasible, sampling too few combinations is meaningless as it will insufficiently interrogate the interface. What then is a logical and reasonable way to reduce the number of variables in such calculations and still arrive at solutions that could achieve novel and improved redesigned protein-protein interfaces? Here, we introduce combYSelect, a computational approach to redesigning protein-protein interfaces that is one way, of many possible ways, to strike a balance between oversampling (i.e., prohibitive computational expense) and undersampling (i.e., insufficient interrogation of the features that contribute to protein-protein interactions).

combYSelect is an *in silico* screening method that relies on restricting the number of possible amino acid mutations at any position to only tyrosine and serine. The choice of these two amino acid options was inspired by previous protein engineering studies showing that phage display libraries of antibody CDR loops with only serine and tyrosine residues gave rise to high-affinity Fabs to a diversity of antigens (26). Our results using combYSelect support the extensive body of literature that describes tyrosine as being beneficial for mediating formation of contacts to promote both affinity and specificity (24). These properties are likely due to the chemical diversity of tyrosine and its ability to form both polar contacts and hydrophobic  $\pi$  interactions. The combination with serine then affords the interface flexibility to form contacts with a variety of residues, as well as additional hydrogen bonding potential. Our results highlight that the properties of tyrosine and serine can also be applied to the redesign of planar interfaces and/or obligate homodimers, *i.e.*, interfaces that have evolved to form highly specific and stable interactions. The flexibility of a Tyr/Ser-populated interface was also evident in the x-ray crystal structures of the combYSelect heterodimers in which electron density of our x-ray crystal structures indicated that the introduced tyrosines occupy multiple conformations. It is possible that the disordered tyrosines contribute to the stability of the interface due to the occurrence of local entropy-enthalpy reinforcement, in which the binding of the two Fc monomers not only is stabilized by favorable enthalpy but also exhibits favorable entropy at the tyrosine interface due to the existence of multiple possible states. This has been previously described on a larger scale as a thermodynamic driver of complex formation between intrinsically disordered proteins (53). We also restricted the number of simultaneous mutations within a single chain to two. Together, these restrictions to the number of variables resulted in 3,459,600 possible variants with which we performed a single Rosetta  $\Delta\Delta G$  calculation. This was completed in approximately 15 hours, a manageable amount of time to comprehensively interrogate a protein-protein interface formed by two Ig-fold subunits.

We applied combYSelect to redesign the interface between the CH3 domains on either protomer in the IgG1 Fc homodimer, with an eye toward developing heterodimeric Fc regions that could serve as platforms for bispecific antibodies (bsAbs). The importance of bsAbs is rapidly growing due to their improved clinical efficacy as compared to monospecific antibodies or combination therapy, as well as for their potential as diagnostic tools (46, 54), particularly for antitumor monoclonal antibodies (mAbs) (55–57). Commonly, bsAbs are engineered to simultaneously target T cell-specific antigens and tumor-associated antigens (TAAs), bridging the two cell types together and enhancing the immune response to tumor cells (58). Another common use of bsAbs is for the targeting of two different TAAs simultaneously, which improves specificity and decreases the likelihood of tumors escaping by down-regulating specific epitopes or reverting to redundant pathways (59).

We describe a method that is highly effective for the design of IgG-like bsAbs, which should retain long serum half-lives due to physiological binding to FcRn, compared to non-IgG-like bsAbs that lack the Fc region (46, 60). To functionally evaluate our combYSelect bsAbs, we replaced one of the Fabs with a nanobody to avoid mismatches between Hcs and Lcs. We did so because Lc association problems have long hindered bsAb development and production. One or both Lcs incorrectly pairing with the Hcs can lead to the formation of unwanted IgGs with aberrant antigen binding capabilities,

thus reducing the yield of the desired bsAb. As a result, similar to the strategies we have previously described to promote proper Hc association, several strategies have been developed to circumvent the Lc pairing problem. Such methods include using a common Lc for both Fab Hcs. However, such chains are difficult to identify and may restrict the diversity of antigens that can be targeted. Other strategies are more similar to Hc heterodimerization methods such as introducing KiH or electrostatic mutations in either the VH-VL interface or the CH1-CL interface. Finally, others have resorted to swapping certain Hc and Lc Fab domains in a strategy known as CrossMab (61). Considering the similarities in folds and interactions between the CH1 domains of both Hc and Lc and the CH3 domains of IgG1, it would be reasonable to suspect that applying combYSelect to promote specific associations between an Lc and its correct Hc pair could lead to the development of a bsAb that most closely resembles the functionality, stability, and long half-life of a WT antibody while retaining the advantages of dual-antigen specificity.

Additionally, combYSelect could potentially be used as a platform for designing bsAbs of other antibody isotypes, considering that conformations and homodimeric interfaces of the Fcs of other antibody isotypes are similar to IgG Fcs. Although all current approved mAb therapeutics are IgGs, progress in the protein engineering field has expanded the possibility of using alternative isotypes such as IgA and IgE. Protein and glyco-engineering efforts have improved the producibility, stability, and half-life of IgA, as monomeric IgA has shown efficacy in preclinical models (62). IgA has unique properties that include its ability to engage neutrophils, which highly express Fc  $\alpha$  receptor I (Fc $\alpha$ RI), making it potentially effective as an antitumor therapeutic. Additionally, IgA displays increased stability in mucosal surfaces compared to IgGs, improving potential for mAbs directed at lung- or gastrointestinal-associated inflammation (63, 64). Similarly, monoclonal IgE binds with extremely high affinity to Fc $\epsilon$  receptor I (Fc $\epsilon$ RI), which is expressed on tumor-associated macrophages, and can improve molecular allergy diagnostics (65, 66). Thus, engineering bispecific IgA and IgE could further contribute to improving the clinical efficacy of such mAbs. A bispecific IgE antibody has been engineered using the KiH strategy, confirming that heterodimeric strategies originally developed for IgGs can be expanded to other antibody isotypes (65).

We developed combYSelect using a library restricted to tyrosine and serine residues motivated by experimental evidence that such a restricted amino acid library could be effective in protein design, albeit for antibody CDR loops binding to antigens. Although we applied it successfully to redesigning the CH3-CH3 interface in IgG1 antibodies, there is no reason that libraries restricted to other combinations of amino acids would not also be useful in protein interface redesign. In the context of structured protein folds, protein-protein interfaces are often hydrophobic, containing a large nonpolar surface area. The formation of these interfaces is largely driven by hydrophobic interactions, such as van der Waals forces (67) and electrostatic forces (68, 69). This merits the consideration of whether different amino acids might be more suitable when redesigning Fcs. A systematic analysis of protein-protein interaction hotspots showed that tryptophan, tyrosine, and arginine are most commonly found in hotspots likely due to the aromatic nature of tryptophan and tyrosine, allowing them to form  $\pi$ -stacking interactions and the ability of all three to form hydrogen bonds. Leucine, serine, threonine, and valine are largely absent from protein hotspots. Additionally, analysis of various protein-protein complexes has found that amino acids

aspartate and asparagine are favored over glutamate and glutamine, and isoleucine is favored over leucine (70). These findings suggest that, within the context of redesign of protein-protein interfaces, amino acids such as arginine, aspartate, asparagine, and isoleucine should also be considered along with tyrosine and serine when deciding the chemical space to be explored.

## MATERIALS AND METHODS

### Residue selection for screening

Residues were selected based on a distance cutoff. An in-house Perl script was used to identify all residues within the homodimeric interface of an AlphaFold model of IgG1 within 4 Å of the opposing chain. Only residues that were located within the CH3 domain of the IgG were used for screening. A total of 31 residues were selected using this method.

### Rosetta interface mode

In silico screening of potential mutations to promote heterodimer formation was carried out using the “interface” mode of Rosetta2.3 (19). Command line options were specified to include extra chi1, chi2, and chi3 rotamers (-extrachi\_cutoff 1 -ex1 -ex2 -ex3). Only the mutant side chains were repacked (the default behavior of this mode), while the protein backbone from the WT structure was retained.

### Cloning

The sequence for Hc and Lc of rituximab was ordered from Thermo Fisher Scientific in the pcDNA3.4-TOPO vector. The Fc plasmid was subcloned out of the Hc plasmid via polymerase chain reaction (PCR). Primers to introduce single or double mutations into the rituximab Hc or Fc plasmids were designed using the Neb-BaseChanger website (<https://nebasechanger.neb.com/>), and PCR was carried out using the NEB High-Fidelity 2X Master Mix and the manufacturer's instructions. All sequences were confirmed by Sanger sequencing at Genewiz (<https://www.genewiz.com/>).

### Recombinant IgG heterodimer expression and purification

All combinations of single-armed IgG proteins were expressed in Expi293 cells. The rituximab Hc, Lc, and Fc plasmids were transfected in a 1:1:1 ratio by weight. The plasmids were transfected as per the manufacturer's protocol (MAN0007814, Thermo Fisher Scientific) with the addition of penicillin/streptomycin mix 24 hours after transfection. The cells were cultured for 96 hours before harvesting. The proteins were purified using protein A resin (Thermo Fisher Scientific), with phosphate-buffered saline (PBS) (pH 7.4) being used as the binding buffer and 100 mM sodium citrate buffer (pH 3.0) as the elution buffer. The fractions were neutralized with 1 M Tris (pH 9.3). The proteins were buffer-exchanged into PBS and stored at 4°C until ready for use. All combinations of potential mutations were tested with the same mutation being tested in both the Hc and Fc plasmids. All proteins were expressed in triplicate in separate transfections.

### Analysis of heterodimer formation

Samples for each of the heterodimer constructs were analyzed via LC-MS to determine the percentage of heterodimer formation. The samples were set up in 96-well plates with a final volume of 20 µl and a concentration of 5 µM. Each expressed protein was analyzed in the absence and presence of EndoS2 (final concentration, 50 nM).

EndoS2 is an IgG-specific endoglycosidase from *Streptococcus pyogenes*, which removes the glycans from the conserved Asn297 site and simplifies data analysis. The samples were analyzed using an Agilent 1290 Infinity II LC System equipped with a 50-mm polymeric reverse-phase column from Agilent with 1000-Å pore size. The LC system is attached to an Agilent 6560 Ion Mobility (IM) Quadrupole Time-Of-Flight mass spectrometer (Agilent, Santa Clara, CA). Relative amounts of each of the deglycosylated peaks were quantified after deconvolution of the raw data and identification of the corresponding peaks using BioConfirm (Agilent, Santa Clara, CA).

### Thermal stability testing

nDSF was carried out using a TychoNT.6 instrument (NanoTemper Technologies, Munich, Germany) to determine the melting temperatures of the monovalent IgG constructs. Samples (20 µl) at 5 µM were loaded into glass capillaries and placed into the sample folder. For unfolding parameters, samples were heated to 95°C at a ramp rate of 30°C/min. Melting temperature  $T_m$  was determined using the instrument software (v1.1.5.668) using first derivative analysis of 350-nm/330-nm fluorescent ratios plotted against temperature. The inflection points that are converted to peaks and troughs are automatically assigned as Tm1/Tm2/Tm3 by the software. All samples were analyzed in triplicate.

### Binding of heterodimer constructs to FcRn

For the two sets of mutations that showed successful heterodimer formation, full-length IgG constructs using only Hc and Lc plasmids were recombinantly expressed and purified in the same manner as above for FcRn binding assays.

BLI was conducted on an OCTETRed384 system at 30°C. A human FcRn-b2m protein construct containing an AviTag was recombinantly expressed in Expi293F cells and biotinylated using BirA. The biotinylated FcRn was loaded onto streptavidin-coated biosensors at a concentration of 10 µg/ml. Immobilization levels between 0.6 and 1.0 nm were reached. The assay buffer was 100 mM phosphate and 150 mM NaCl (pH 6).

For association phase monitoring, IgG samples were diluted with assay buffer (pH 6) with a starting concentration of 1000 nM. Samples (40 µl) composed of seven threefold dilutions were made and transferred into OCTET 384 tilted well plates. The eighth sample was left at 0 nM to provide a baseline measurement. The IgG samples were allowed to bind to the FcRn-loaded biosensors for 300 s. The dissociation phase was recorded in a solid black 96-well plate for 150 s. The biosensors were regenerated between runs for 30 s with 10 mM Hepes (pH 8). The regeneration steps were carried out twice, with the sensors washed with pH 6 buffer for 30 s. The experiments were carried out the same way at neutral pH, except the pH of the assay buffer was adjusted to pH 7.4. All experiments were carried out in triplicate.

All data were referenced with the FcRn-loaded streptavidin-coated biosensors loaded only in assay buffer. The sensograms were collected using the ForteBio Data Acquisition software v11.1 and plotted and analyzed using steady-state fits with the ForteBio Data Analysis Software v 11.1.

### Crystallization of combYSelect1 and combYSelect2 Fcs

combYSelect1 and combYSelect2 Fcs were expressed as described above and concentrated to 5 mg/ml. For crystal generation, the reservoir of 48-3 Intelliplates (Art Robbins 102-0003-00) was filled with 150 µl of solution, and 0.5 µl of protein was combined with

0.5  $\mu\text{l}$  of mother liquor. combYSelect1 Fcs were crystallized in 0.1 M bis-tris (pH 6.5) and 21% PEG MME 5000 (Hampton Research). The crystals were harvested in mother liquor supplemented with 20% (v/v) glycerol for cryoprotection and flash-frozen in liquid nitrogen. combYSelect2 Fcs were crystallized in 0.1 M bis-tris (pH 6.5) and 29% PEG MME 2000 (Hampton Research). The crystals were harvested in mother liquor supplemented with 30% (v/v) glycerol for cryoprotection and flash-frozen in liquid nitrogen.

Data were collected at Southeast Regional Collaborative Access Team (SER-CAT) 22-ID beamline at the Advanced Photon Source, Argonne National Laboratory on Eiger16 detector and processed on HKL2000. Both structures were solved with molecular replacement with an IgG1 Fc [Protein Data Bank (PDB) ID: 7LBL] as a search model using PHENIX Phaser-MR (71). The structures were built and refined using Coot (72) and phenix.refine (73). The models were also further refined with the PDB-REDO server (74).

### Cell-bridging assay

For the cell staining assay, highlighting the potential utility of bsAbs, the previously characterized 5F7 nanobody, which is specific for HER2, was used. The nanobody string was purchased from Twist Bioscience and subcloned into the Hc vector replacing the CH1 domain of the Lc. Like the heterodimers, these constructs were transfected in a 1:1:1 ratio and purified as above.

BT474 and Raji cells were resuspended to a density of  $1 \times 10^7$  cells/ml in  $1 \times$  PBS. The BT474 cells were stained with Calcein-Violet-AM (Thermo Fisher Scientific, catalog no. C34858) and the Raji cells were stained with CellTrace CFSE (carboxyfluorescein diacetate succinimidyl ester) (Thermo Fisher Scientific, catalog no. C34554) according to the manufacturer's protocols, centrifuged, and resuspended in staining buffer (BD Biosciences, catalog no. 554656). The cells were combined in equal volume/density, and human BD Fc block (catalog no. 564220) was added according to the manufacturer's protocol. The combined cells (200  $\mu\text{l}$ ) were added to wells of a 96-well flat bottom plate (Corning, catalog no. 3917). The cells were then stained with primary antibody (50  $\mu\text{g}/\text{ml}$ ) for 45 min on ice, washed three times with staining buffer, and resuspended in the same buffer before collecting data on an Agilent NovoCytte Penton flow cytometer. This protocol was adapted from a previously published version of this assay (65).

### IL-1 cytokine trap inhibition assay

HEK293T-derived cells grown in serum-free F17 medium [American Type Culture Collection (ATCC), ACS-4500] were transiently transfected with a luciferase gene under the control of the IL-8 promoter (nano-luc, plasmid pNL2.2, Promega) using FugeneHD (Promega). After 18 hours, cells were harvested and seeded into 96-well plates at a concentration of 40,000 cells per well. Tenfold serial dilutions of the antagonist (electrostatics trap, combYSelect2 trap, or IL-1Ra) were prepared with the concentrations ranging from 0.1  $\mu\text{M}$  to 0.1 pM. The cells were pretreated with the antagonist for 15 min and stimulated with 5 pM IL-1 $\beta$ . Cells that were stimulated with cytokine in the absence of antagonists and cells that did not receive any stimulation were used as controls. After a 5-hour incubation at 37°C, the cells were lysed and the luciferase activity was determined using a BioTek luminescence reader. The activity of cells stimulated with cytokine in the absence of the antagonist was used to normalize luminescence data. This protocol was adapted from a previously published version of this assay (48, 75).

### Statistical analysis

All statistical analysis was done in GraphPad Prism. Ordinary one-way analyses of variance (ANOVAs) were used for comparison of multiple groups with one independent variable, followed by Tukey's multiple-comparison tests when comparing each mean with all other means or Dunnett's multiple-comparison tests when comparing each mean with one control mean. \* $P < 0.05$ , \*\* $P < 0.01$ , \*\*\* $P < 0.001$ , and \*\*\*\* $P < 0.0001$ . For the IL-1 cytokine trap inhibition assay, the percentage of luciferase response was normalized to that when no inhibitor is added. The IC<sub>50</sub> (median inhibitory concentration) values were determined using a nonlinear least square fit. All intact LC-MS analysis, FcRn binding, thermal stability testing, and IL-1 cytokine trap inhibition assay were performed in independent triplicates.

### Supplementary Materials

This PDF file includes:

Figs. S1 to S5

Table S1

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