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## 5 Prediction of Antibody Non-Specificity using Protein 6 Language Models and Biophysical Parameters

7 **Supporting Information**

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16

## 18 List of Figures

Figure	Description
S1	Distribution of non-specificity from three public antibody datasets: (A) Boughter dataset, (B) UMAP projection of the sequence similarity distances of the H/L-CDRs for the antibodies within the Boughter dataset, (C) boxplot of H-CDR3 length of the antibodies within the Boughter dataset, (D) Jain dataset, (E) Shehata dataset, and (F) balance of non-specificity class within individual datasets.
S2	10-fold CV for different antibody domain input embedded by ESM 1v: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S3	10-fold CV for different antibody domain input embedded by ESM 1b: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S4	10-fold CV for different antibody domain input embedded by ESM 2: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S5	10-fold CV for different antibody domain input embedded by Protbert bfd: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S6	10-fold CV for different antibody domain input embedded by AntiBERTy: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S7	10-fold CV for different antibody domain input embedded by AbLang2: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S8	10-fold CV for different antibody domain input embedded by 68 different sequence-based descriptors: panel (A) Accuracy, panel (B) Sensitivity, and panel (C) Specificity.
S9	Spearman correlation matrix of 68 VH-based sequence-based descriptors.
S10	Bar plots showcasing accuracy of models with top descriptor combinations and their frequency among the top 15 models: panel (A) top 2-descriptor models (top 15 out of 300 combinations), panel (B) top 3-descriptor models (top 15 out of 2300 combinations), panel (C) top 4-descriptor models (top 15 out of 12650 combinations), and panel (D) top 5-descriptor models (top 15 out of 53130 combinations). Each bar represents a different combination of 2-5 descriptors, with error bars indicating the standard deviation of 10-fold CV accuracy. The red line and red shaded area indicate the 10-fold CV accuracy and standard deviation, respectively, for ESM 1v VH-based LogisticReg model.
S11	Visualisation of feature importance (absolute value of Eigenvalues) for 5 components from the PCA of VH sequence descriptors.
S12	VH-based LogisticReg models showcasing validation performance (k-Fold CV and Leave-One-Family-Out) for PLM- and descriptor-based models: (A) Sensitivity and (B) Specificity. Bar plot for accuracy can be found in Figure 2C in main paper.
S13	Boxplot showing the predicted non-specificity probabilities for the respective ELISA flag of the Jain dataset using VH-based Logistic Regression (ESM 1v). The boxplot displays the median, interquartile range, and outliers, with strength of regression-like trend indicated by SCC and p-value (<0.001).
S14	Confusion matrices for VH-based LogisticReg models across different datasets showcasing the number of antibodies correctly and incorrectly predicted as specific (label 0) and non-specific (label 1). Each panel represents a different model and dataset combination: panel (A) ESM 1v VH-based LogisticReg model tested on the Jain dataset, panel (B) Top 5-descriptors VH-based LogisticReg model tested on the Jain dataset, panel (C) ESM 1v VH-based LogisticReg model tested on the Shehata dataset, panel (D) Top 5-descriptors VH-based LogisticReg model tested on the Shehata dataset, panel (E) ESM 1v VH-based LogisticReg model tested on the Harvey dataset, and panel (F) Top 5-descriptors VH-based LogisticReg model tested on the Harvey dataset.

S15	Distribution of top VH-based sequence descriptors for specific and non-specific antibodies in the Boughter dataset (mildly non-specific antibodies excluded). Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.
S16	Distribution of top VH-based sequence descriptors for specific and non-specific antibodies in the Jain dataset (mildly non-specific antibodies excluded). Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.
S17	Distribution of top VH-based sequence descriptors for specific and non-specific antibodies in the Shehata dataset. Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.
S18	Distribution of top VH-based sequence descriptors for specific and non-specific antibodies in the Harvey dataset. Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.
S19	Performance of Harvey <i>et al.</i> predictor (2022) on various datasets: panels (A-B) Boughter dataset, panels (C-D) Jain dataset, and panels (E-F) Shehata dataset.

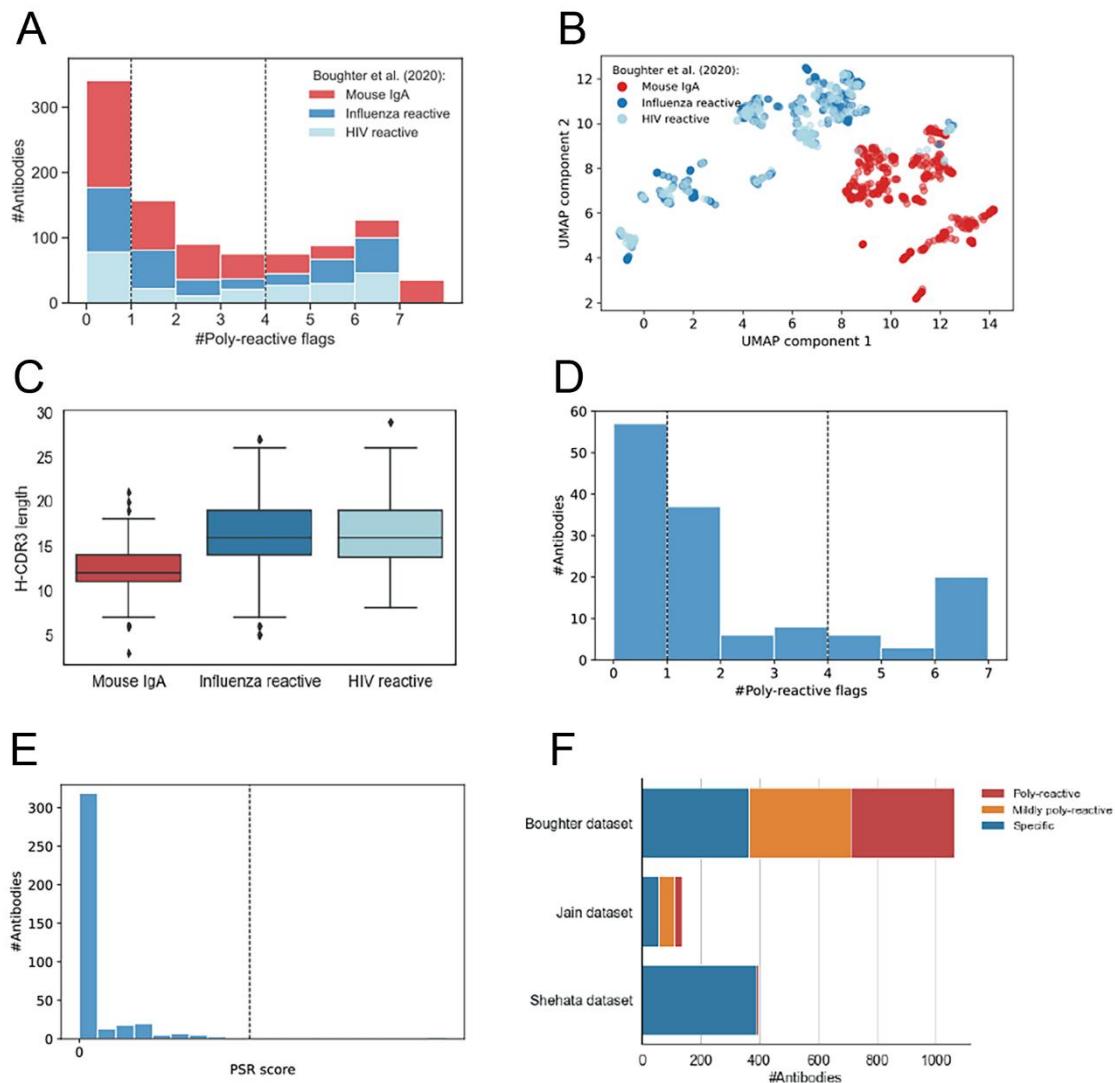
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## 21 List of Tables

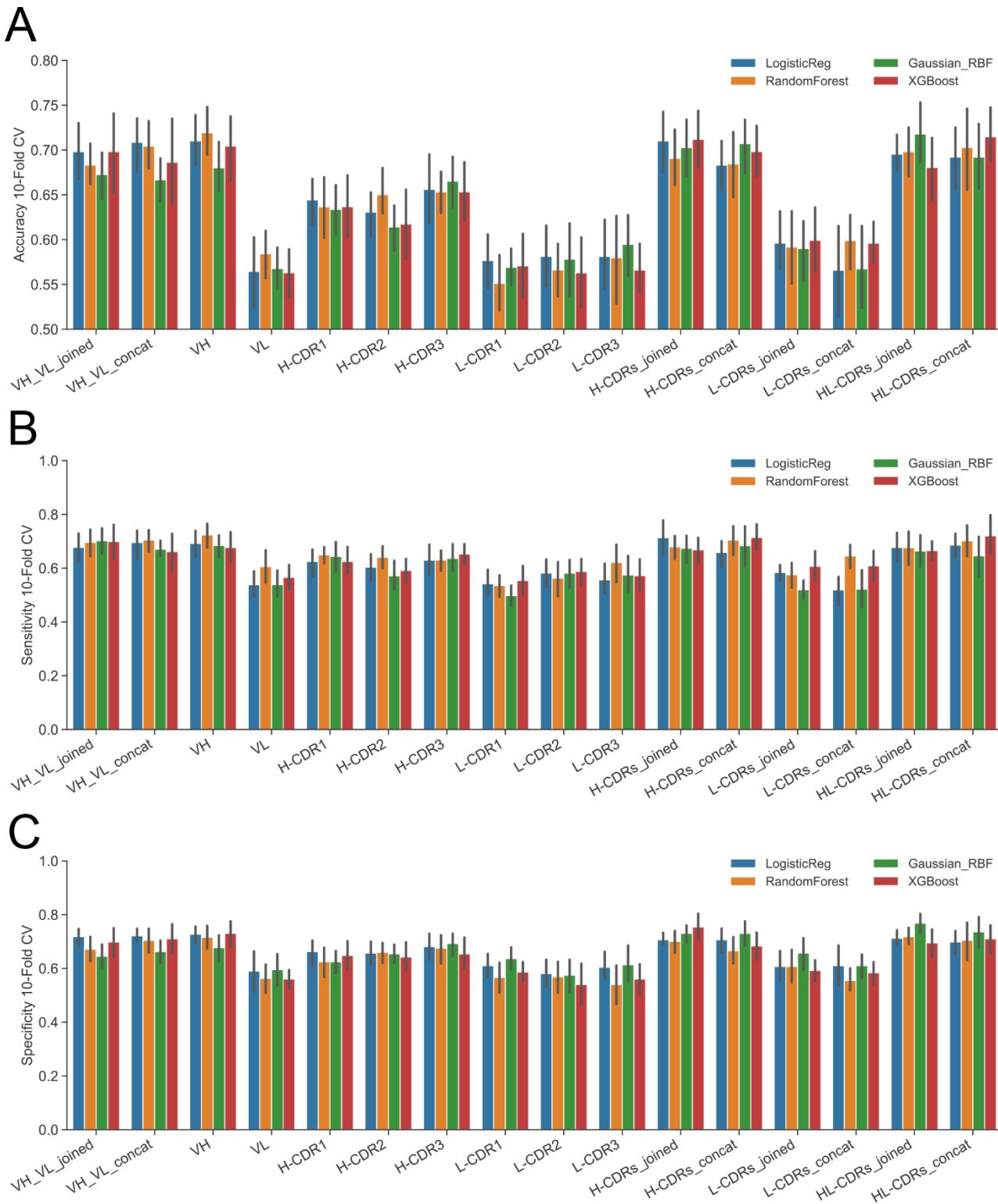
Table	Description
S1	Overview of biophysical descriptors. All the descriptors are derived from Schrödinger but those marked (*), which have been calculated with the Biopython ProteinAnalysis module. Further documentation on the Schrödinger descriptors can be found at <a href="https://support.schrodinger.com/s/article/827119">https://support.schrodinger.com/s/article/827119</a> .
S2	Summary of descriptor importance and model performance for a VH-based LogisticReg model trained on all descriptors (excluding charge at pH 6 and 7.4). For each descriptor, the following is provided; (i) cluster number assigned by hierarchical clustering of Spearman correlation coefficients as a way to group redundant descriptors, (ii) LogisticReg coefficients from the logistic regression model, (iii) the decrease in model accuracy when the descriptor is permuted, indicating its importance (based on 10-fold CV on test data), (iv) the model accuracy when the specific descriptor is left out, indicating its unique contribution, and (v) the accuracy of the model using only the single descriptor. Descriptors are listed along with their respective cluster number and their importance metrics, highlighting their contribution to the model's performance and their individual predictive power.

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24 **Figure S1.** Distribution of non-specificity from three public antibody datasets: (A) Boughter  
 25 dataset, (B) UMAP projection of the sequence similarity distances of the H/L-CDRs for the  
 26 antibodies within the Boughter dataset, (C) boxplot of H-CDR3 length of the antibodies within the  
 27 Boughter dataset, (D) Jain dataset, (E) Shehata dataset, and (F) balance of non-specificity class  
 28 within individual datasets.  
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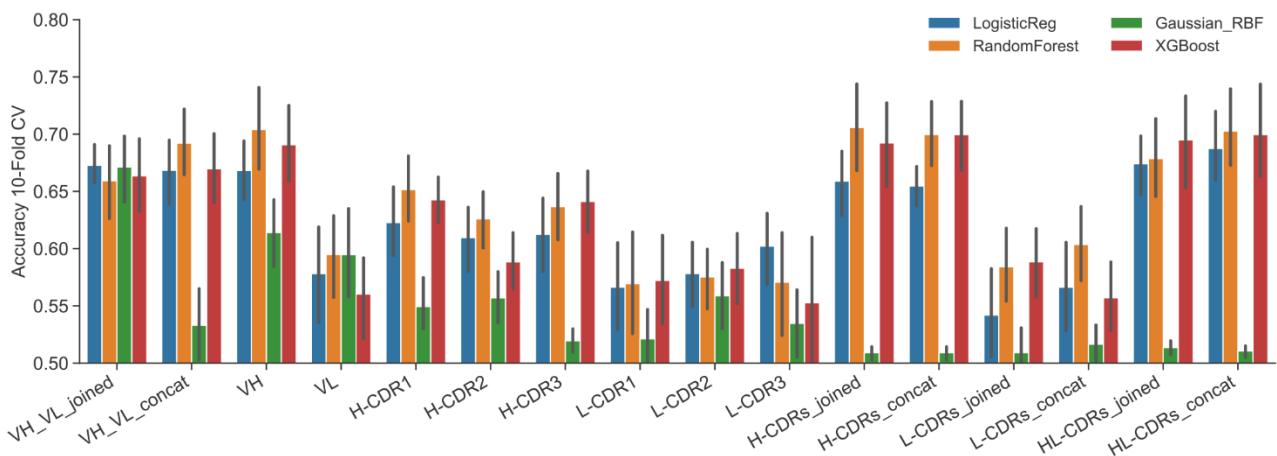


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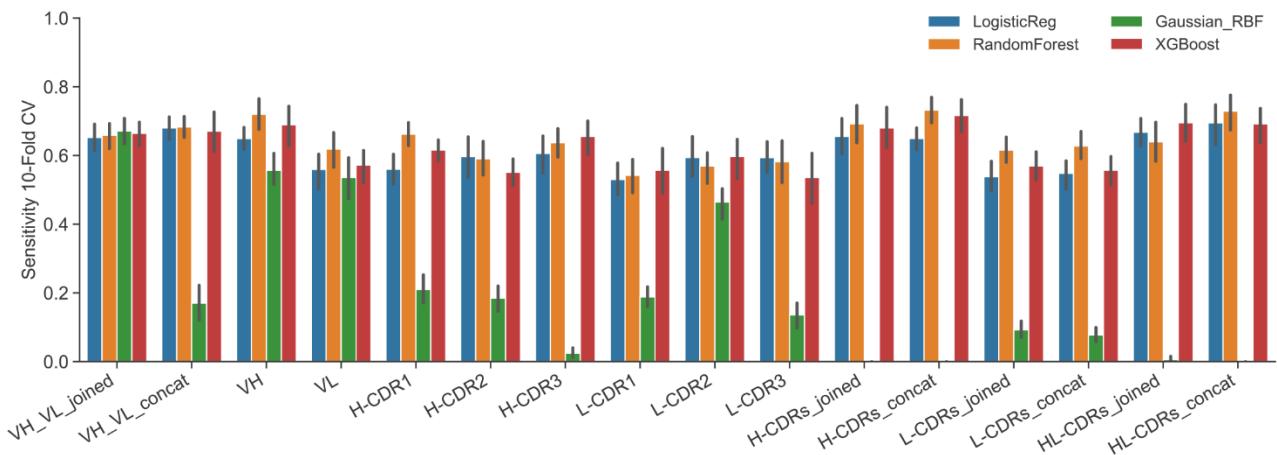
31 **Figure S2.** 10-fold CV for different antibody domain input embedded by ESM 1v: (A) Accuracy,  
32 (B) Sensitivity, and (C) Specificity.

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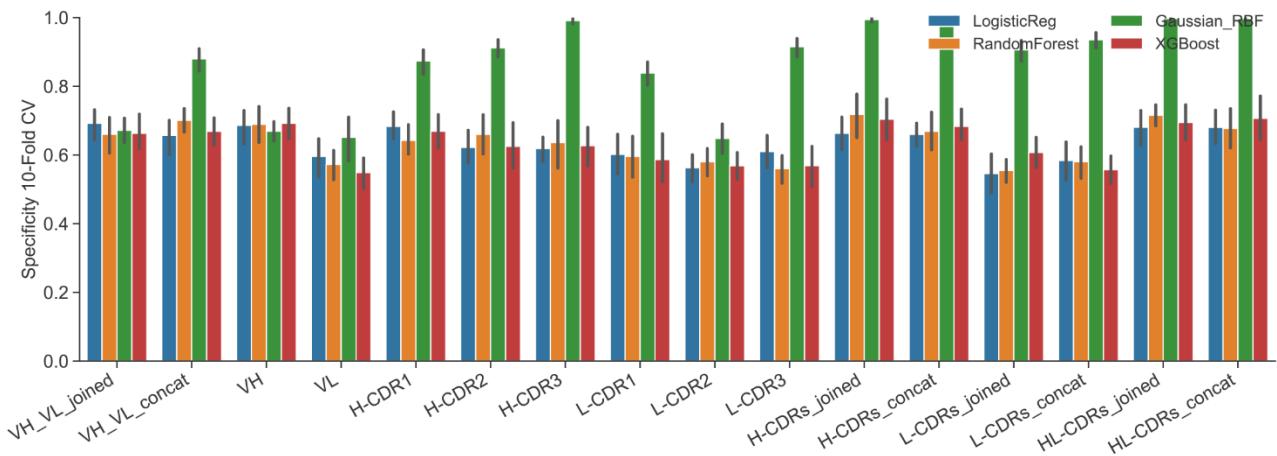
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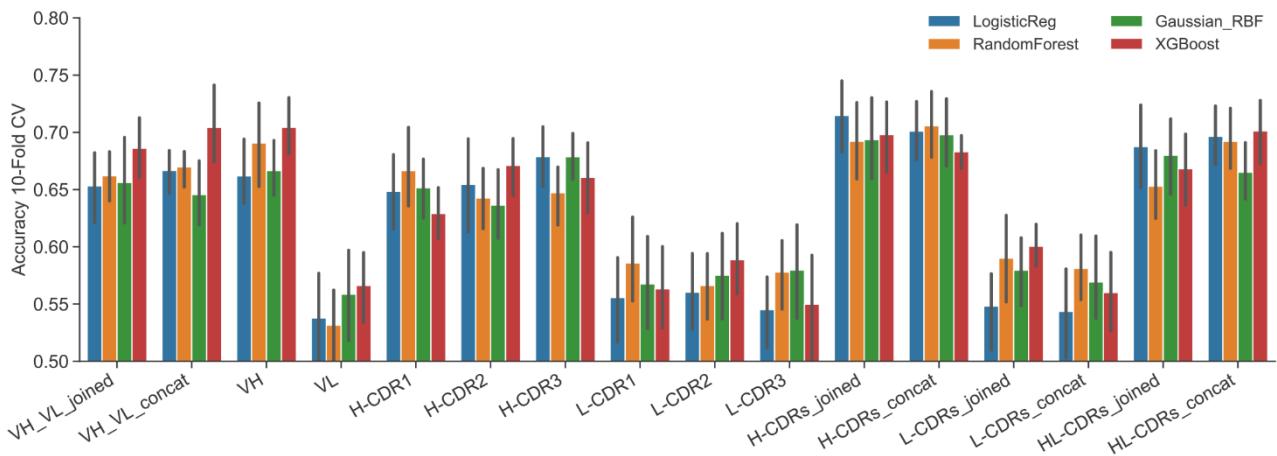
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35 **Figure S3.** 10-fold CV for different antibody domain input embedded by ESM 1b: (A) Accuracy,  
 36 (B) Sensitivity, and (C) Specificity.

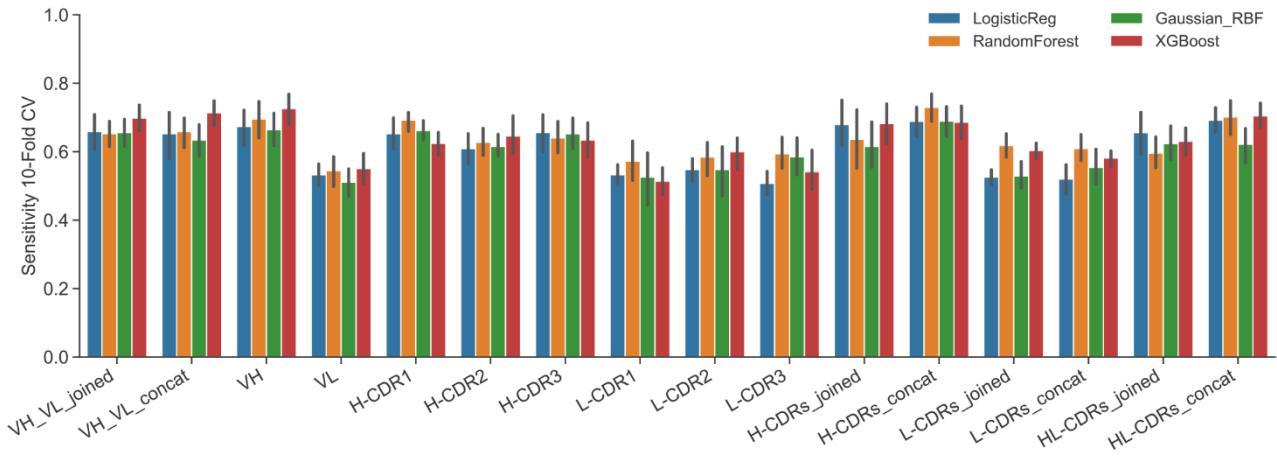
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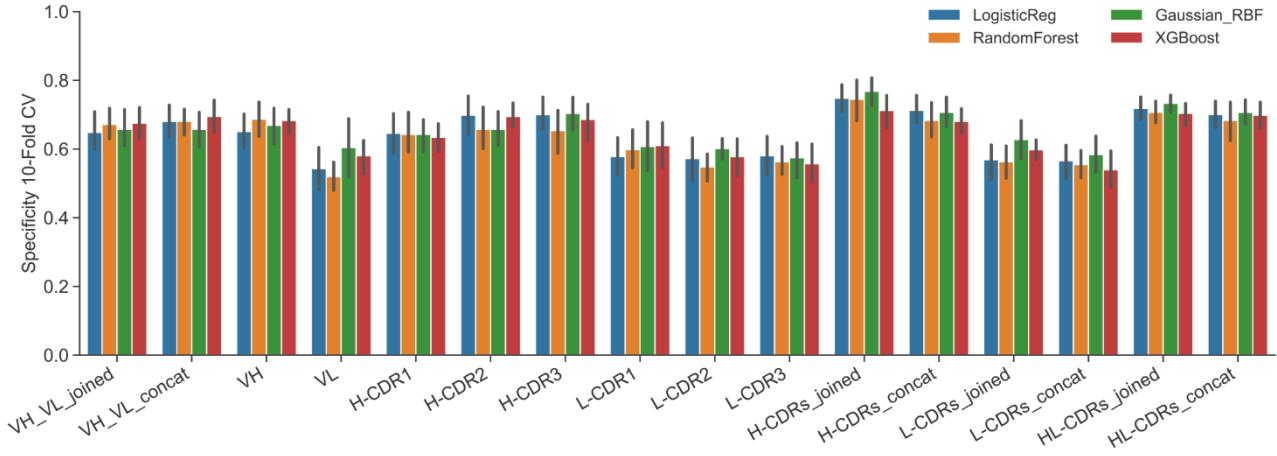
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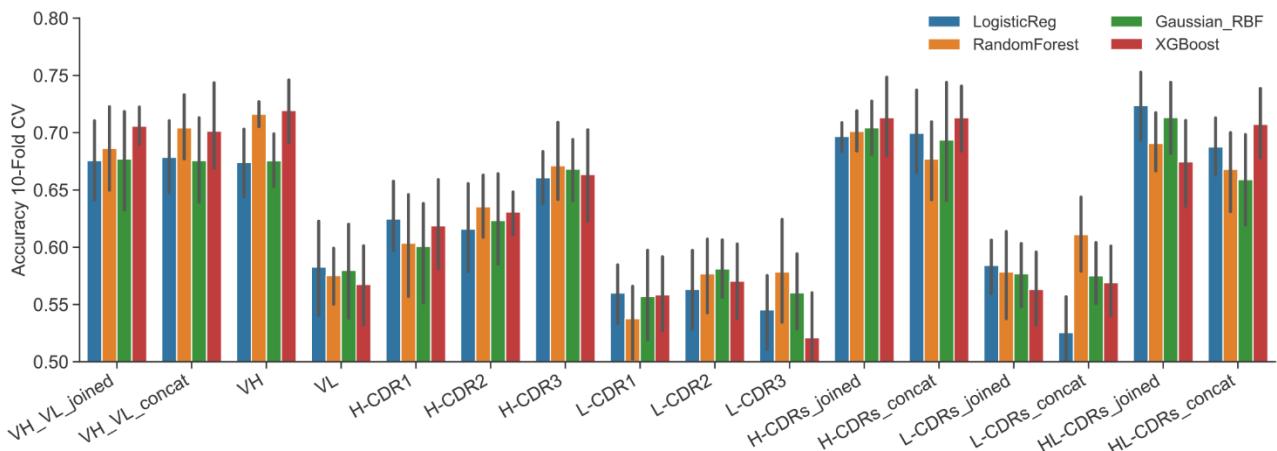


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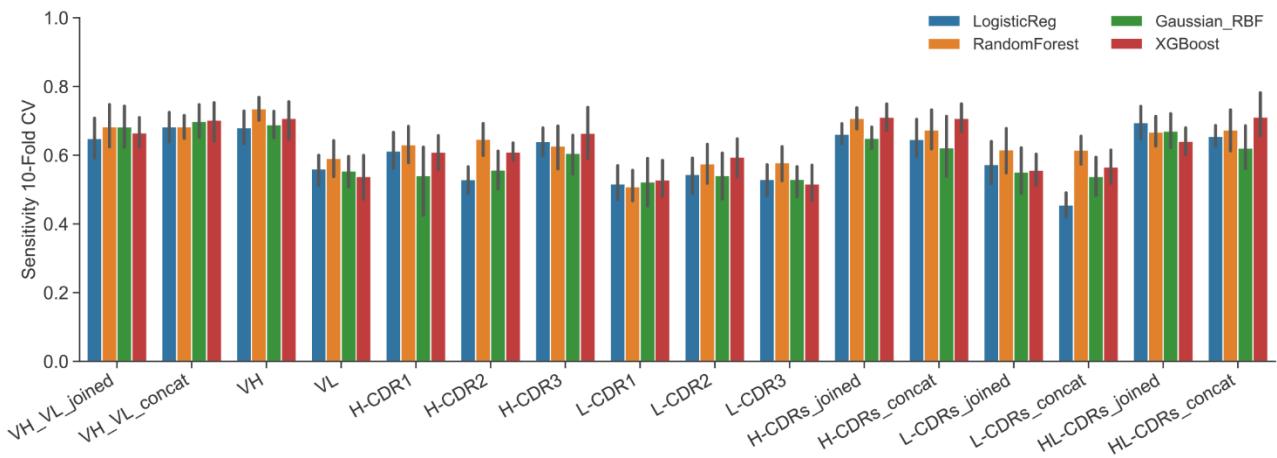
40 **Figure S4.** 10-fold CV for different antibody domain input embedded by ESM 2: (A) Accuracy, (B)  
41 Sensitivity, and (C) Specificity.

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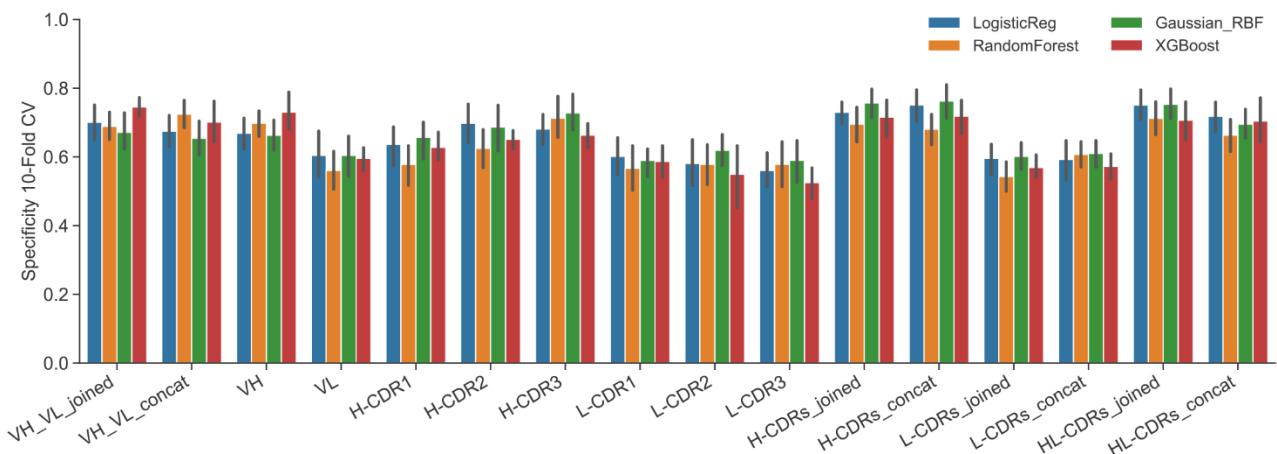
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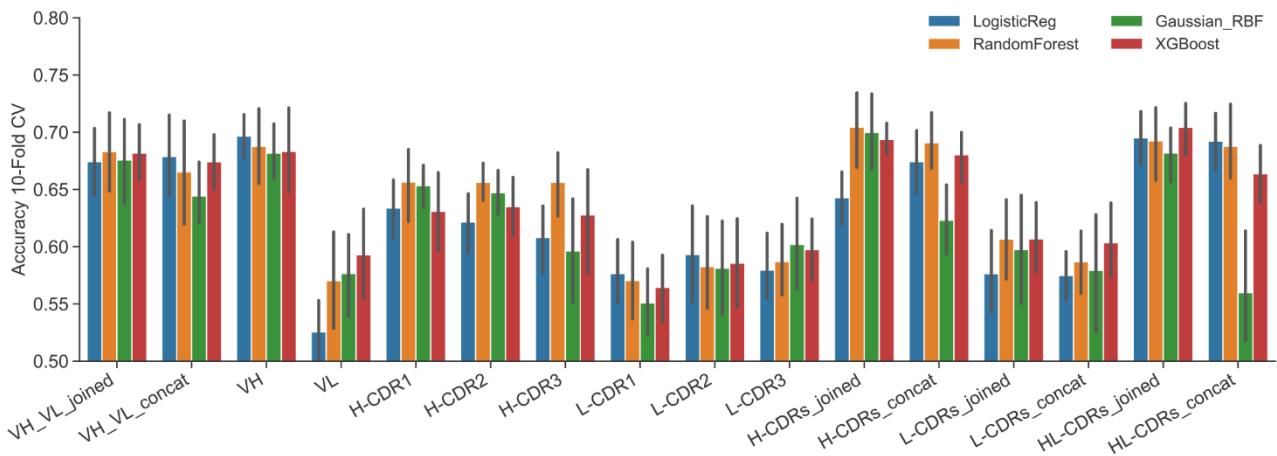


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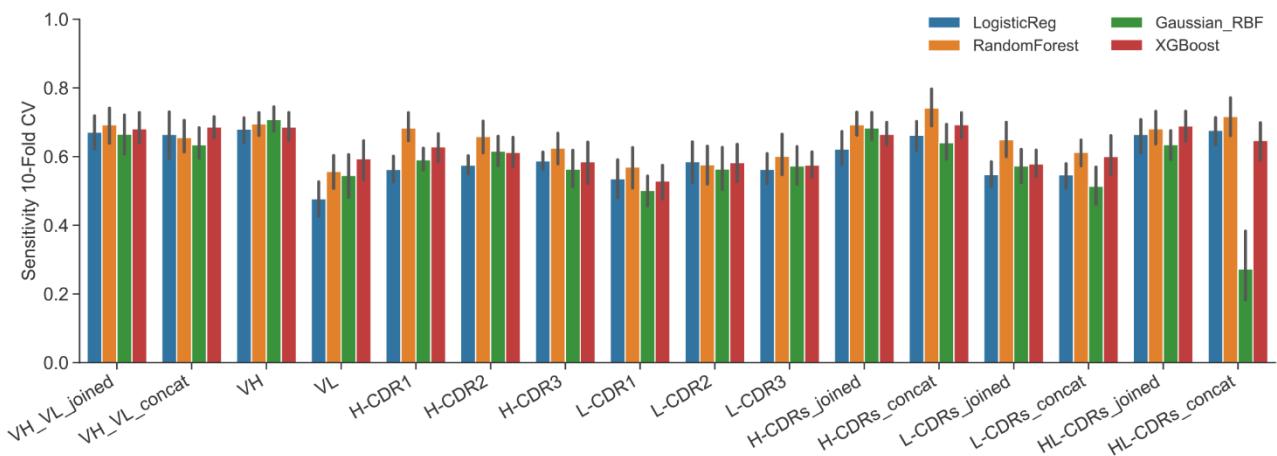
44 **Figure S5.** 10-fold CV for different antibody domain input embedded by Protbert bfd: (A)  
45 (B) Sensitivity, and (C) Specificity.

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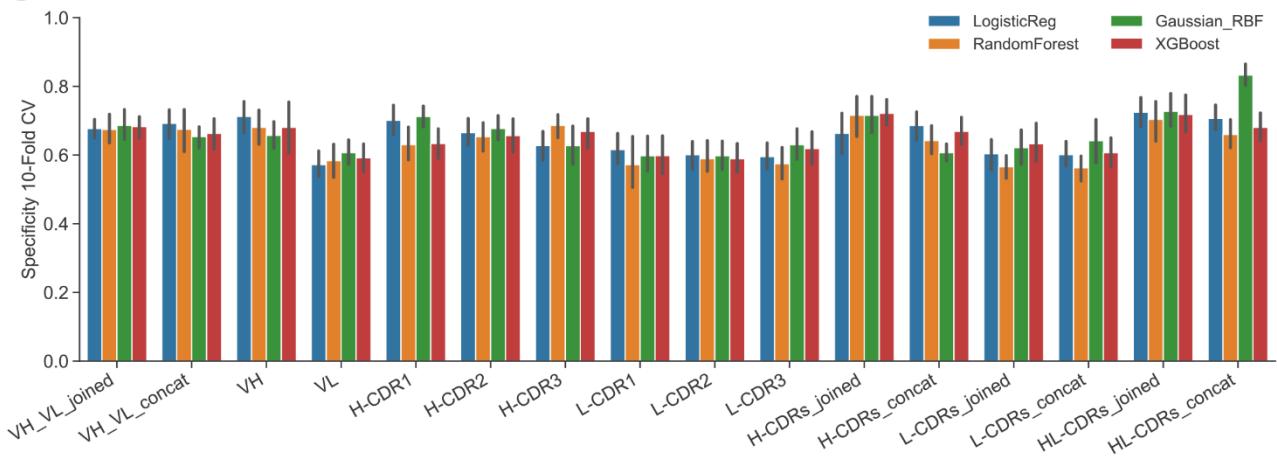
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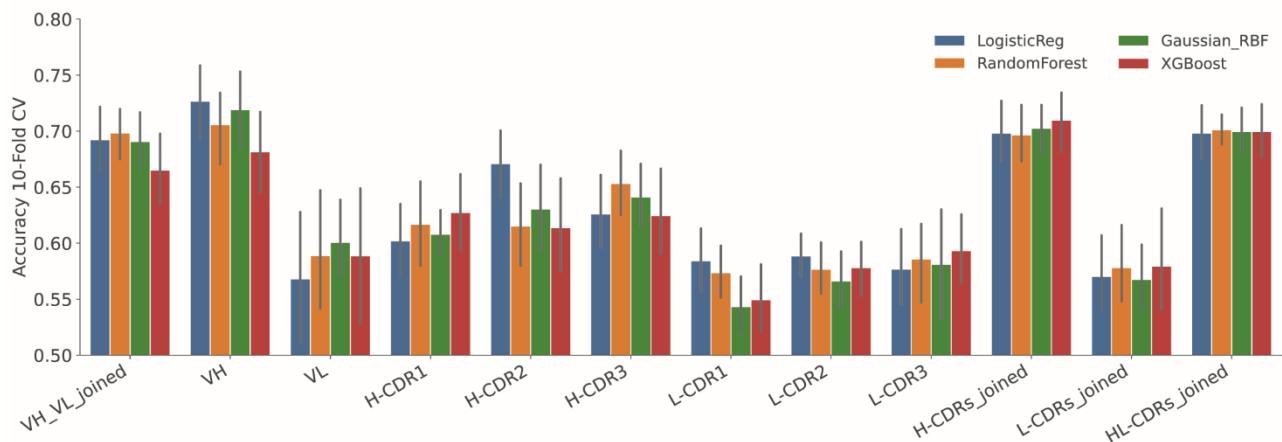
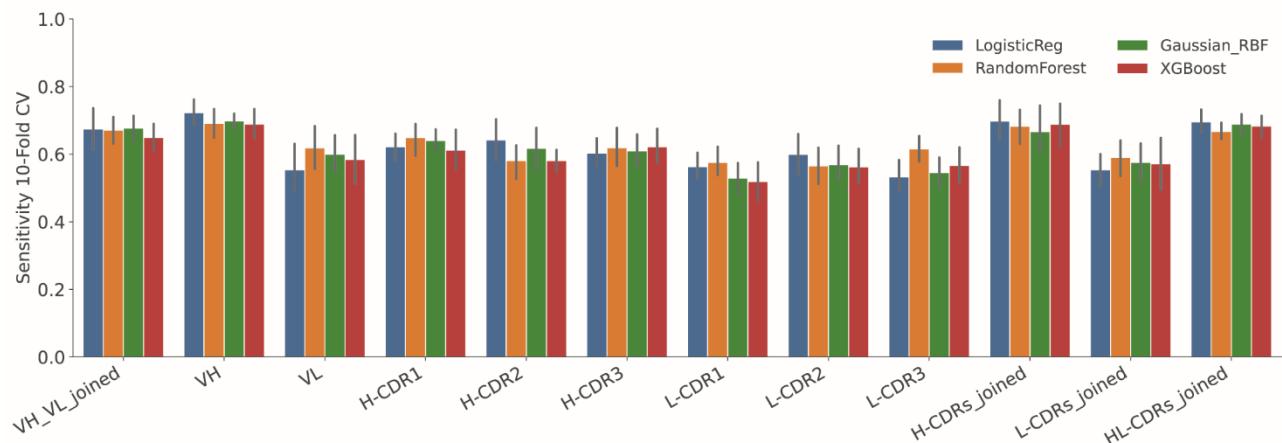
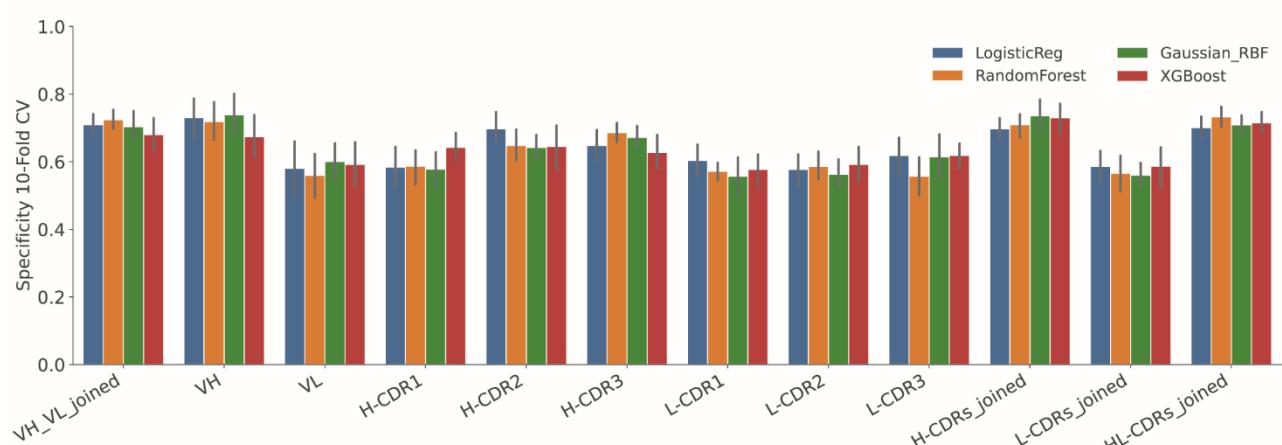


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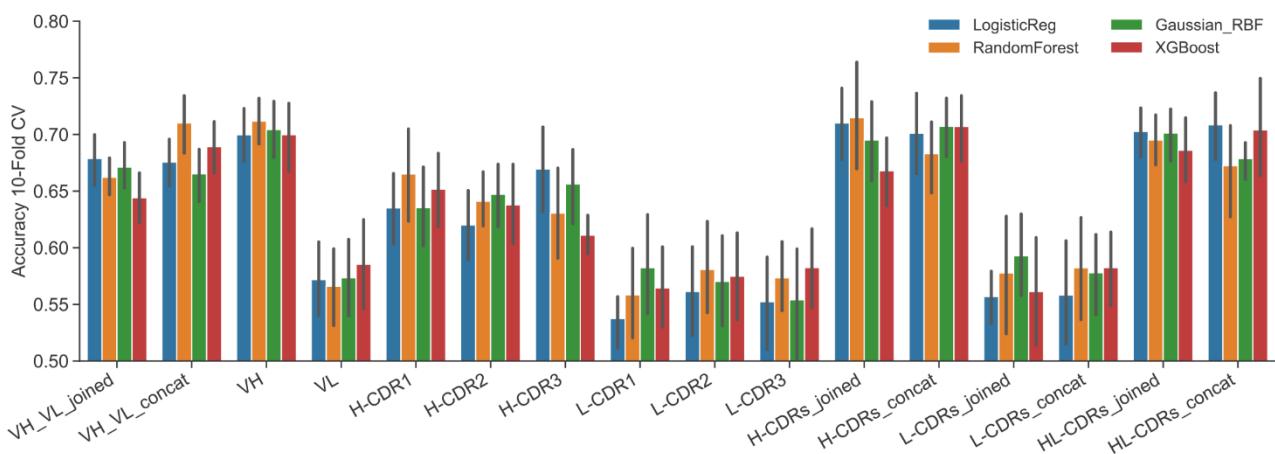
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49 **Figure S6.** 10-fold CV for different antibody domain input embedded by AntiBERTy: (A)  
50 Accuracy, (B) Sensitivity, and (C) Specificity.

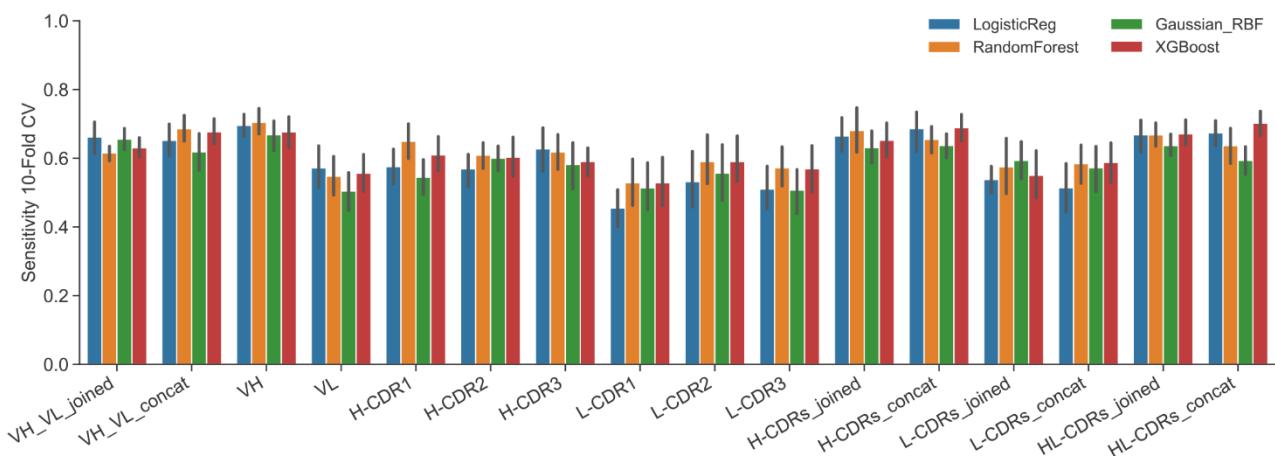
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**A****B****C**

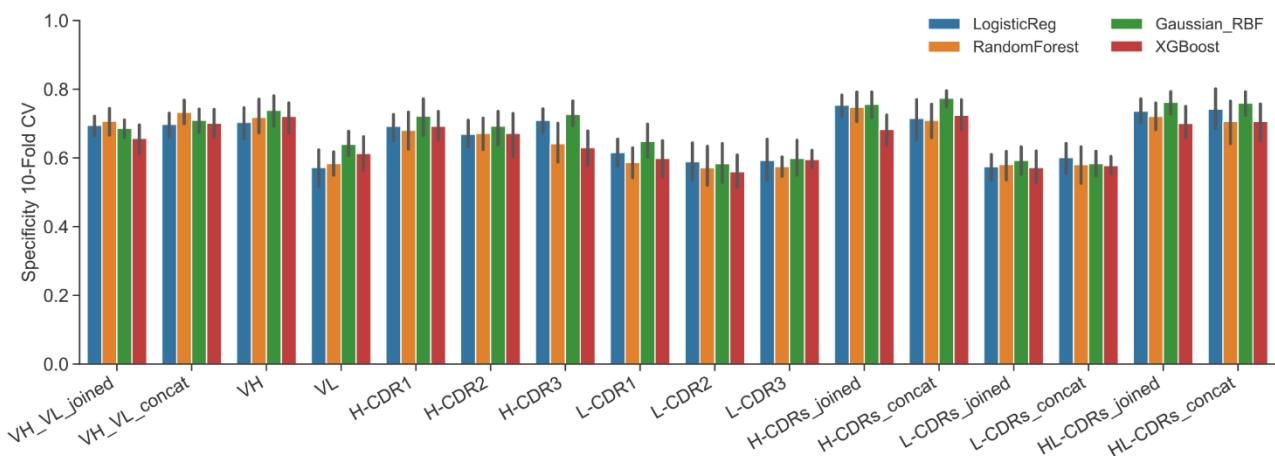
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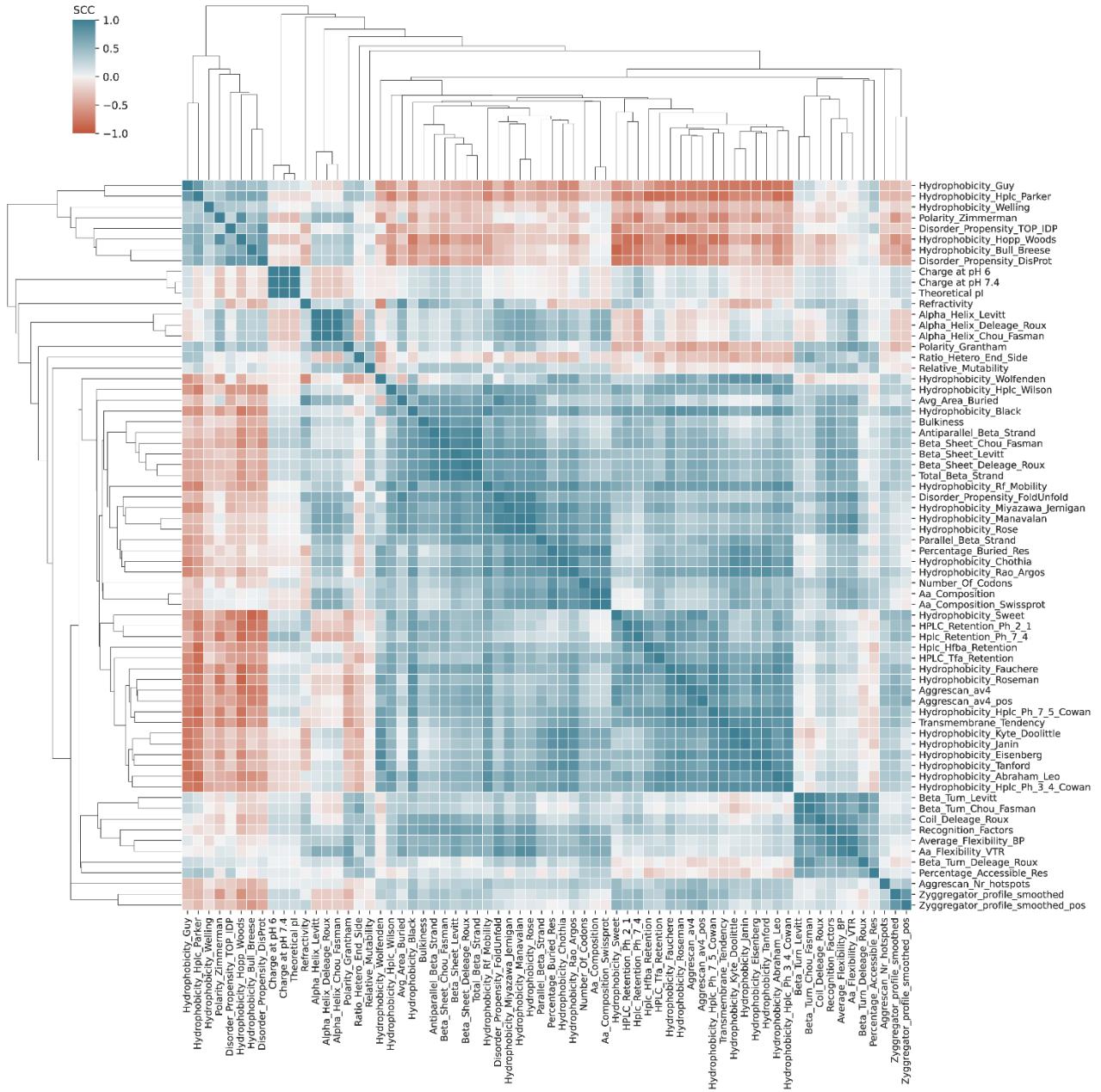


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58 **Figure S8.** 10-fold CV for different antibody domain input embedded by 68 different sequence-based descriptors: (A) Accuracy, (B) Sensitivity, and (C) Specificity.

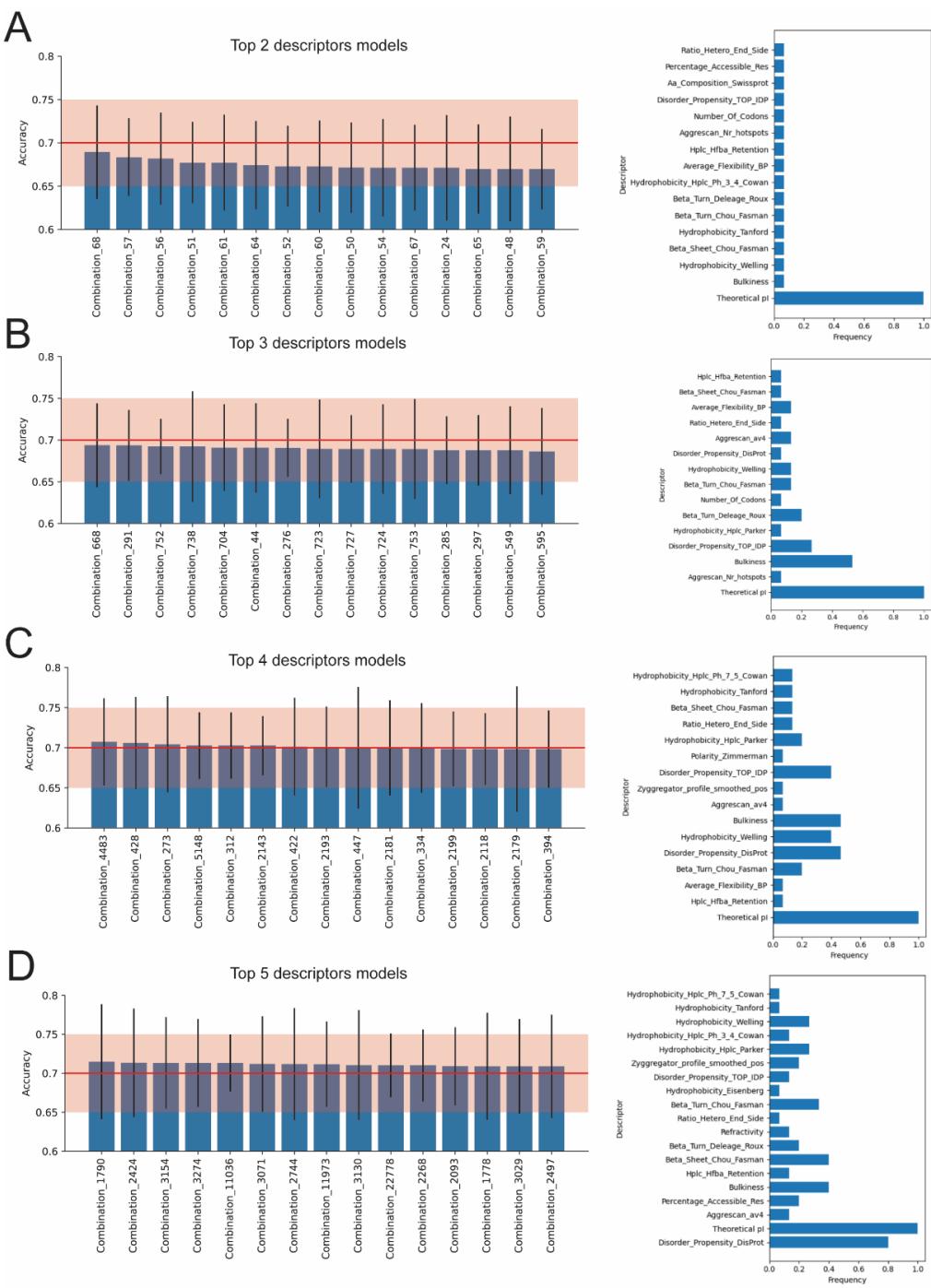
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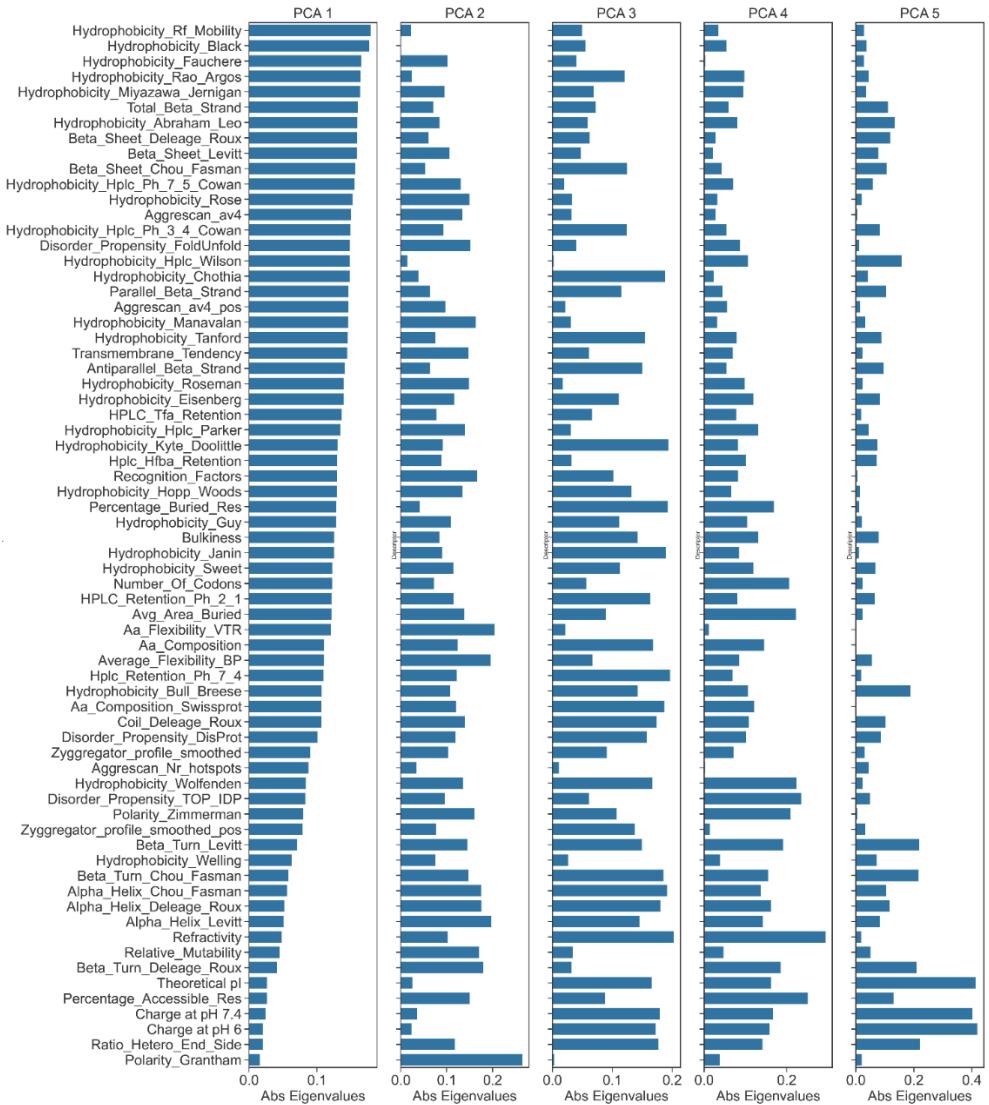
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62 **Figure S9.** Spearman correlation matrix of 68 VH-based sequence-based descriptors.



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64 **Figure S10.** Bar plots showcasing accuracy of models with top descriptor combinations and their  
65 frequency among the top 15 models: (A) top 2-descriptor models (top 15 out of 300 combinations),  
66 (B) top 3-descriptor models (top 15 out of 2300 combinations), (C) top 4-descriptor models (top 15  
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68 Each bar represents a different combination of 2-5 descriptors, with error bars indicating the  
69 standard deviation of 10-fold CV accuracy. The red line and red shaded area indicate the 10-fold  
70 CV accuracy and standard deviation, respectively, for ESM 1v VH-based LogisticReg model.

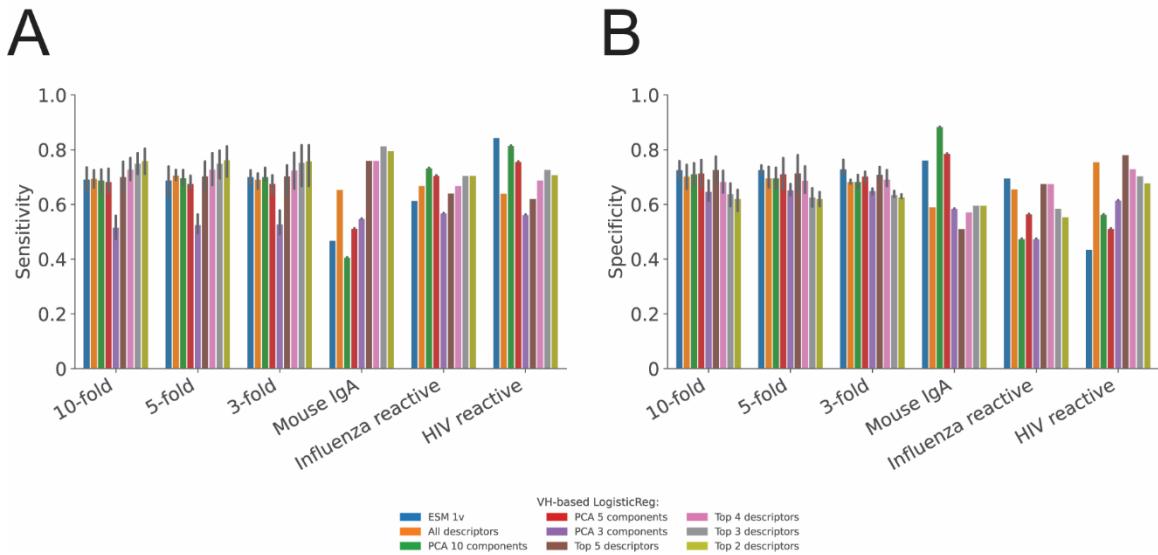


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72 **Figure S11.** Visualisation of feature importance (absolute value of Eigenvalues) for 5 components  
73 from the PCA of VH sequence descriptors.

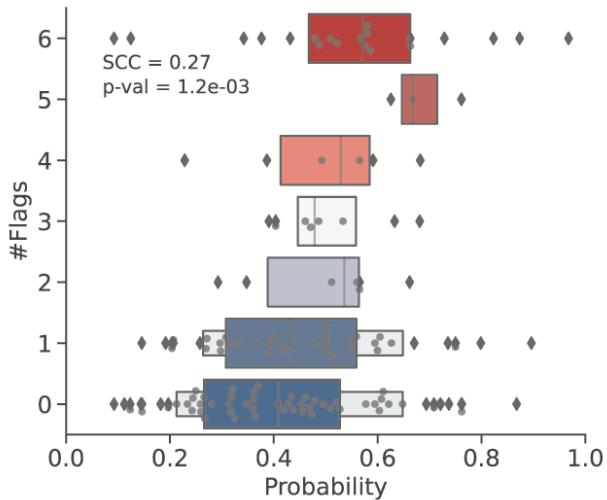
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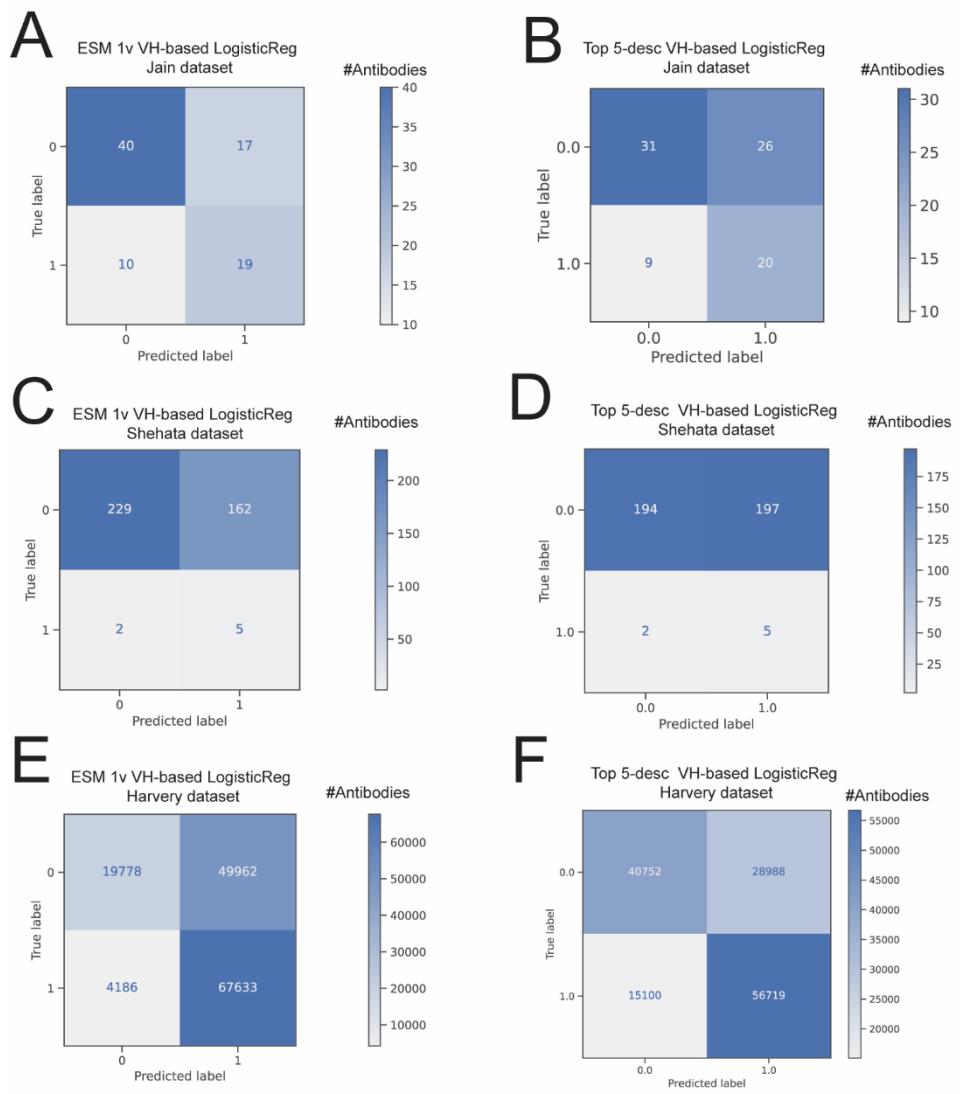
77 **Figure S12.** VH-based LogisticReg models showcasing validation performance (k-Fold CV and  
 78 Leave-One-Family-Out) for PLM- and descriptor-based models: (A) Sensitivity and (B) Specificity.  
 79 Bar plot for accuracy can be found in Figure 2C in main paper.  
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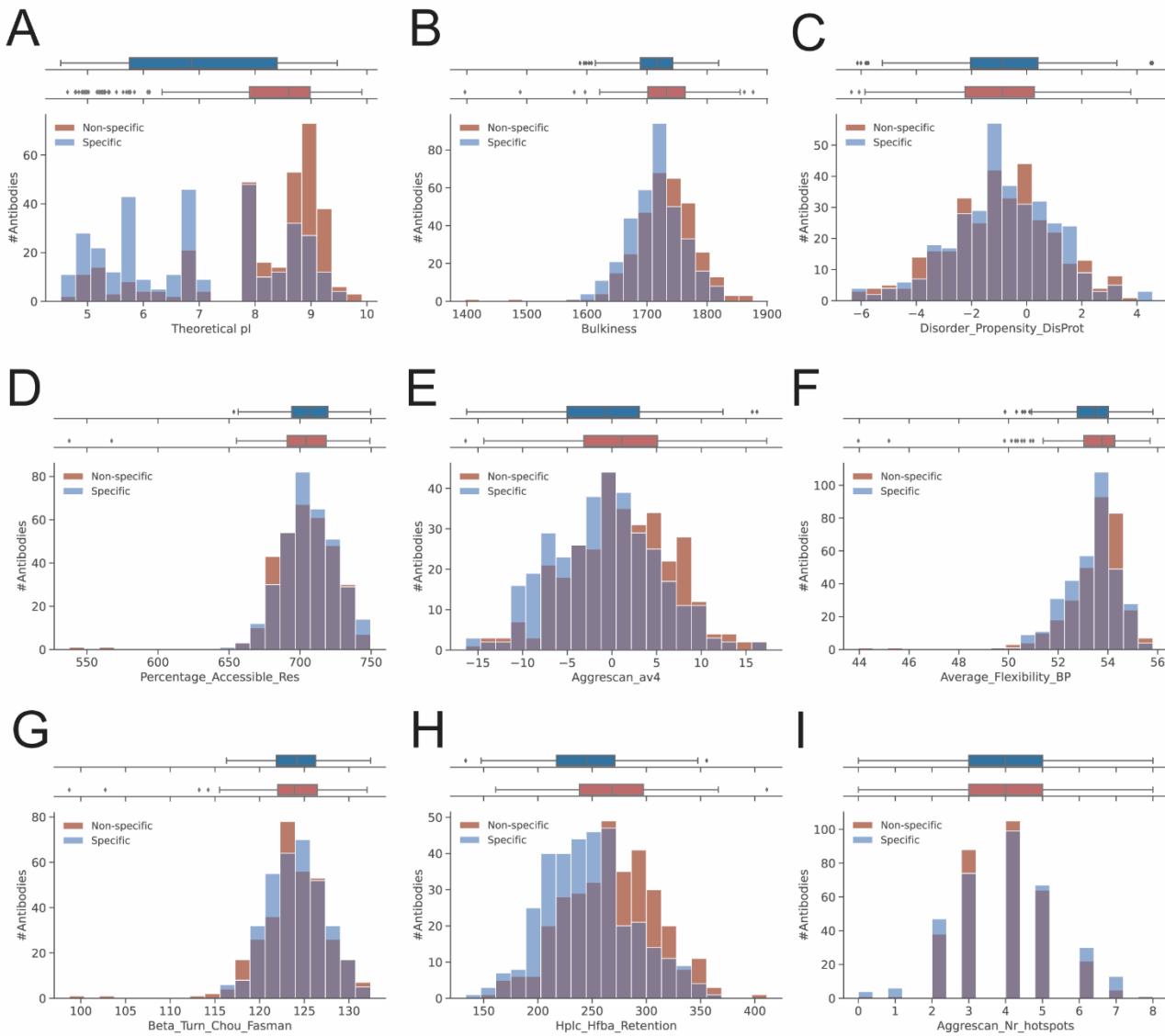
82 **Figure S13.** Boxplot showing the predicted non-specificity probabilities for the respective ELISA  
 83 flag of the Jain dataset using VH-based Logistic Regression (ESM 1v). The boxplot displays the  
 84 median, interquartile range, and outliers, with strength of regression-like trend indicated by SCC  
 85 and p-value (<0.001).

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88 **Figure S14.** Confusion matrices for VH-based LogisticReg models across different datasets  
 89 showcasing the number of antibodies correctly and incorrectly predicted as specific (label 0) and  
 90 non-specific (label 1). Each panel represents a different model and dataset combination: panel (A)  
 91 ESM 1v VH-based LogisticReg model tested on the Jain dataset, panel (B) Top 5-descriptors VH-  
 92 based LogisticReg model tested on the Jain dataset, panel (C) ESM 1v VH-based LogisticReg  
 93 model tested on the Shehata dataset, panel (D) Top 5-descriptors VH-based LogisticReg model  
 94 tested on the Shehata dataset, panel (E) ESM 1v VH-based LogisticReg model tested on the Harvey  
 95 dataset, and panel (F) Top 5-descriptors VH-based LogisticReg model tested on the Harvey dataset.  
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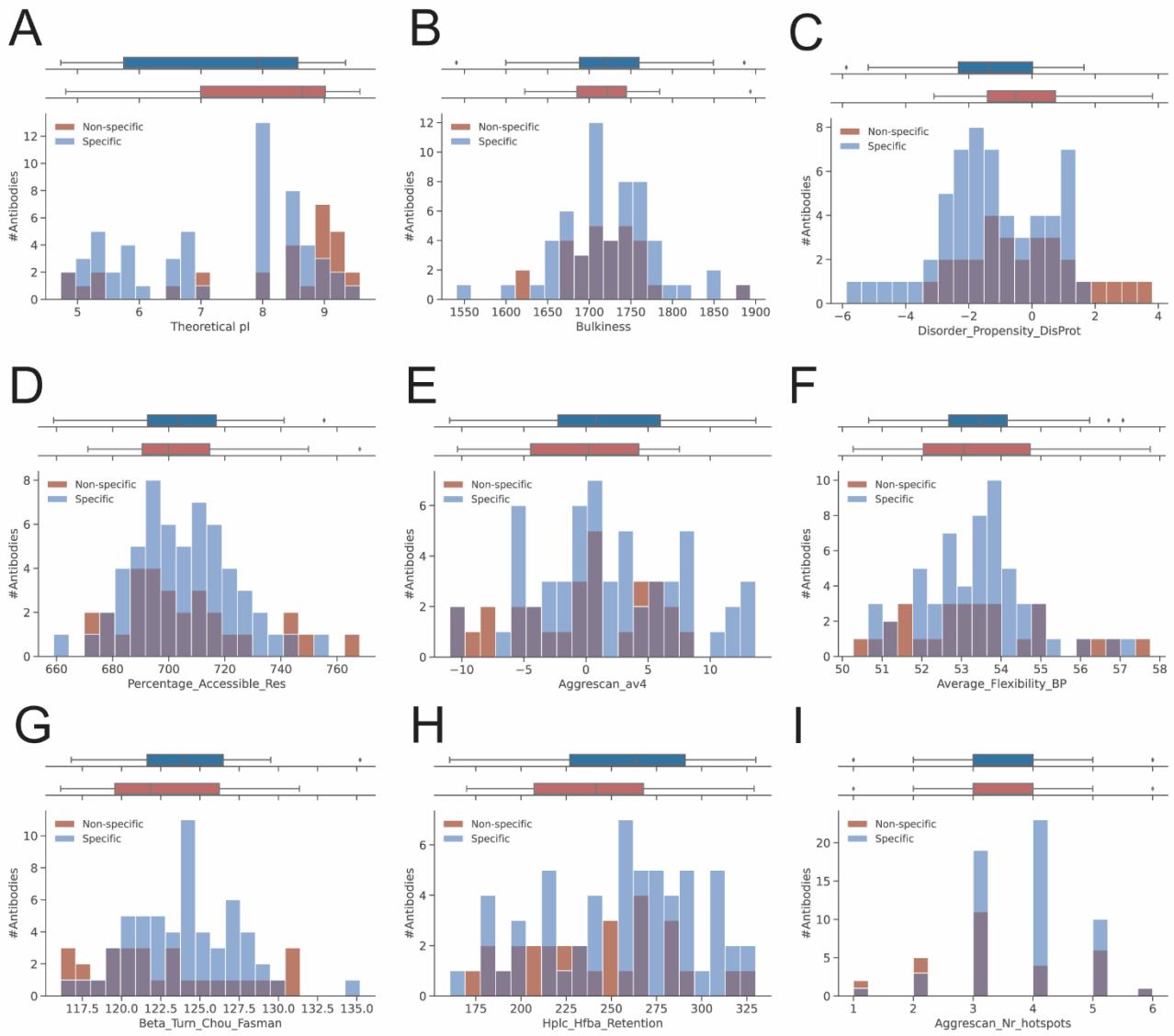
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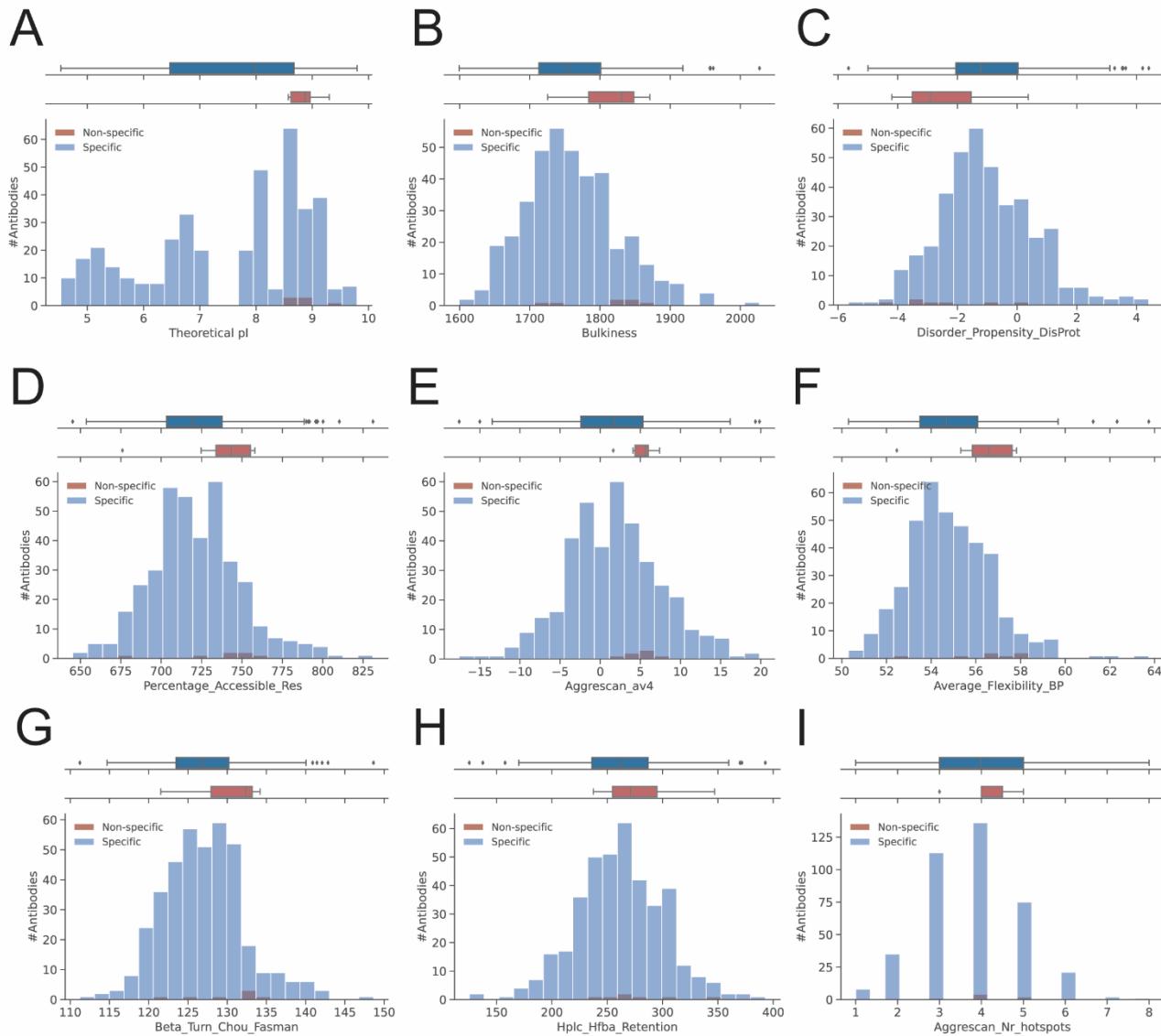
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**Figure S15. Boughter dataset.** Distribution of top VH-based sequence descriptors for specific and non-specific antibodies (mildly non-specific antibodies excluded). Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggrescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggrescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.



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**Figure S16. Jain dataset.** Distribution of top VH-based sequence descriptors for specific and non-specific antibodies (mildly non-specific antibodies excluded). Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.



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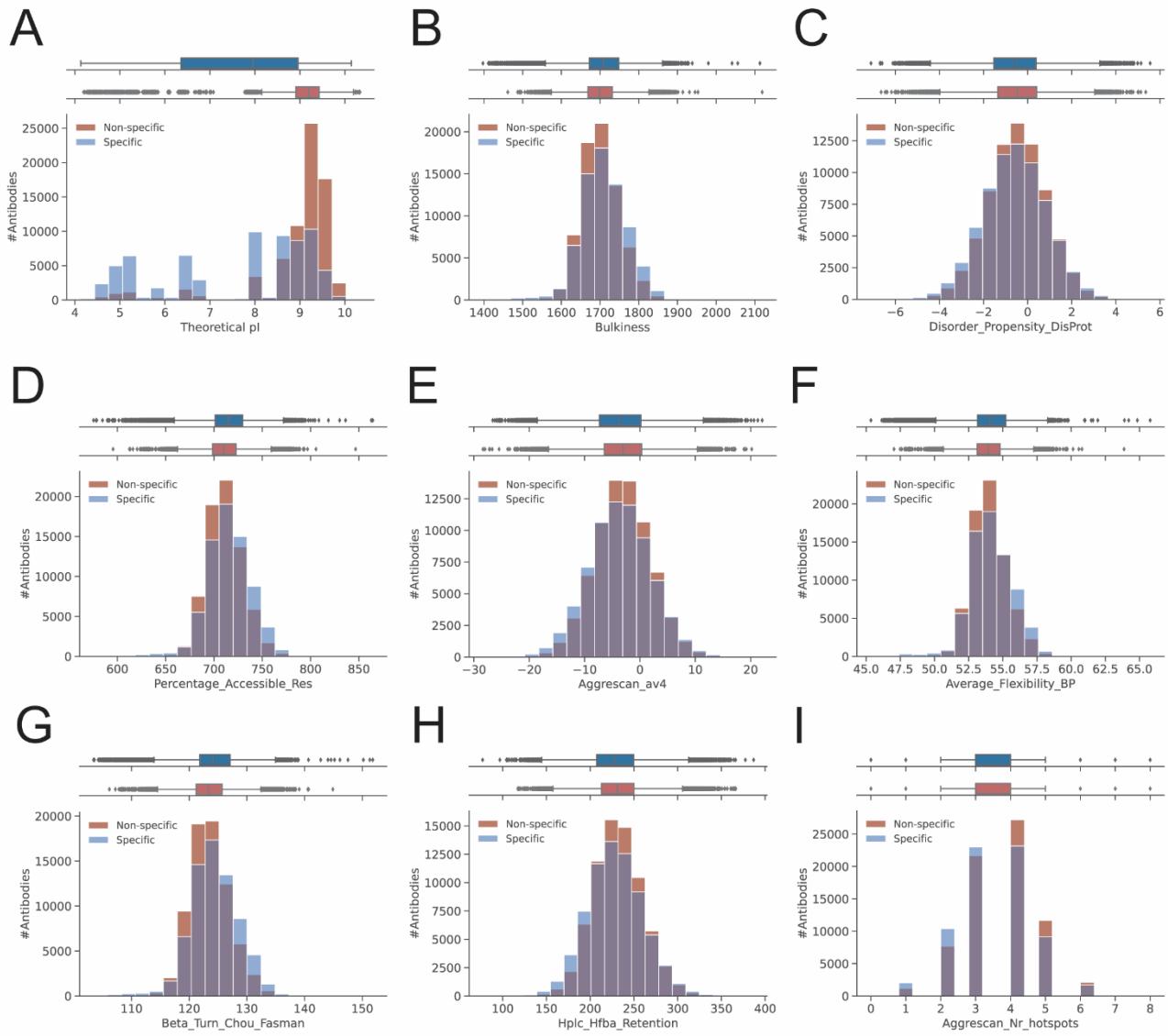
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**Figure S17. Shehata dataset.** Distribution of top VH-based sequence descriptors for specific and non-specific antibodies. Histograms and boxplots show the distribution of various descriptors for specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor: panel (A) theoretical pI, panel (B) bulkiness, panel (C) disorder propensity according to DisProt, panel (D) percentage of accessible residues, panel (E) Aggrescan av4, panel (F) average flexibility using BP scale, panel (G) beta turn propensity according to Chou-Fasman, panel (H) high-performance liquid chromatography retention using HFBA, and panel (I) number of hotspots according to Aggrescan. For each descriptor, the top boxplots display the distribution for specific and non-specific antibodies, with the median and interquartile range indicated. The histograms illustrate the frequency of antibodies within different value bins, providing insights into the characteristics of each descriptor.

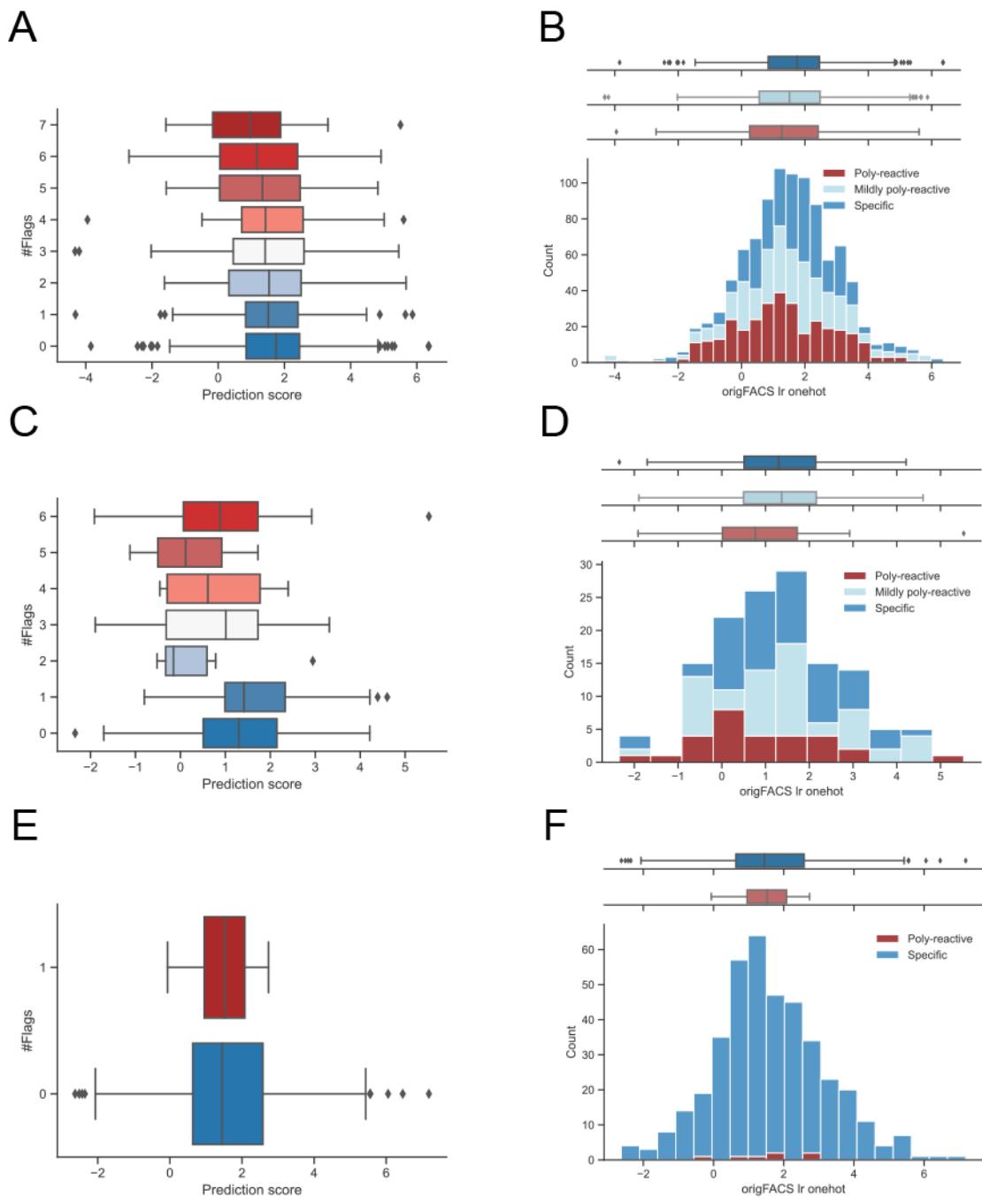


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139 **Figure S18. Harvey dataset.** Distribution of top VH-based sequence descriptors for specific and  
 140 non-specific antibodies. Histograms and boxplots show the distribution of various descriptors for  
 141 specific (blue) and non-specific (red) antibodies. Each subplot represents a different descriptor:  
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 145 performance liquid chromatography retention using HFBA, and panel (I) number of hotspots  
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 148 illustrate the frequency of antibodies within different value bins, providing insights into the  
 149 characteristics of each descriptor.

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153 **Figure S19.** Performance of Harvey *et al.* predictor (2022): (A-B) Boughter dataset, (C-D) Jain  
154 dataset, and (E-F) Shehata dataset.

155      **Table S1.** Overview of biophysical descriptors. All the descriptors are derived from Schrödinger but  
 156      those marked (\*), which have been calculated with the Biopython ProteinAnalysis module. Further  
 157      documentation on the Schrödinger descriptors can be found at:  
 158      <https://support.schrodinger.com/s/article/827119>.

#	Descriptor	Definition
1	AGGRESCAN_Nr_hotspots	Number of aggregation hotspots computed by the Aggrescan algorithm ( <a href="http://bioinf.uab.es/aap/aap_help.html">http://bioinf.uab.es/aap/aap_help.html</a> )
2	Aa_Composition	The total of amino acid composition as described by McCaldon and Argos (Proteins: Structure, Function and Genetics 4:99-122(1988))
3	Aa_Composition_Swissprot	The total value of amino acid composition based on SwissProt annotation (Release notes for UniProtKB/Swiss-Prot release 2013_04 - April 2013)
4	Aa_Flexibility_VTR	The total amino acid flexibility as defined by Vihtinen, Torkkila, and Rikonen ( <a href="https://www.ncbi.nlm.nih.gov/pubmed/8090708">https://www.ncbi.nlm.nih.gov/pubmed/8090708</a> )
5	Aggrescan_av4	a4v values over a sliding window, as determined by the Aggrescan algorithm
6	Aggrescan_av4_pos	a4v positive values over a sliding window, as determined by the Aggrescan algorithm
7	All_Aggrescan_a4v_pos	The sum of the average of a4v positive values over a sliding window, as determined by the Aggrescan algorithm
8	Alpha_Helix_Chou_Fasman	Alpha helix propensity, as defined by Chou and Fasman (Adv. Enzym. 47:45-148(1978))
9	Alpha_Helix_Deleage_Roux	Alpha helix propensity, as defined by Deleage and Roux (Protein Engineering 1:289-294(1987))
10	Alpha_Helix_Levitt	Alpha helix propensity, as defined by Levitt (Biochemistry 17:4277-4285(1978))
11	Antiparallel_Beta_Strand	Antiparallel beta strand propensity, as defined by Lifson and Sander (Nature 282:109-111(1979))
12	Average_Flexibility_BP	Total amino acid flexibility, as defined by Bhaskaran and Ponnusamy (Int. J. Pept. Protein. Res. 32:242-255(1988))
13	Avg_Area_Buried	Average standard-state to folded-protein buried area, as defined by Rose et al. (Science 229:834-838(1985))
14	Beta_Sheet_Chou_Fasman	Beta sheet propensity, as defined by Chou and Fasman (Adv. Enzym. 47:45-148(1978))
15	Beta_Sheet_Deleage_Roux	Beta sheet propensity, as defined by Deleage and Roux (Protein Engineering 1:289-294(1987))
16	Beta_Sheet_Levitt	Beta sheet propensity, as defined by Levitt (Biochemistry 17:4277-4285(1978))
17	Beta_Turn_Chou_Fasman	Beta turn propensity, as defined by Chou and Fasman (Adv. Enzym. 47:45-148(1978))
18	Beta_Turn_Deleage_Roux	Beta turn propensity, as defined by Deleage and Roux (Protein Engineering 1:289-294(1987))
19	Beta_Turn_Levitt	Beta turn propensity, as defined by Levitt (Biochemistry 17:4277-4285(1978))
20	Bulkiness	Total amino acid bulkiness (J. Theor. Biol. 21:170-201(1968))
21	Charge at pH 6*	Charge of the protein at pH 6
22	Charge at pH 7.4*	Charge of the protein at pH 7.4
23	Coil_Deleage_Roux	Total score for coil, as defined by Deleage and Roux (Protein Engineering 1:289-294(1987))
24	Disorder_Propensity_DisProt	Total disorder promotion propensity ( <a href="https://www.ncbi.nlm.nih.gov/pubmed/17578581">https://www.ncbi.nlm.nih.gov/pubmed/17578581</a> )
25	Disorder_Propensity_FoldUnfold	Total disorder promotion propensity ( <a href="https://www.ncbi.nlm.nih.gov/pubmed/15498936">https://www.ncbi.nlm.nih.gov/pubmed/15498936</a> )
26	Disorder_Propensity_TOP_IDP	Total disorder propensity for intrinsic disorder, based on the TOP-IDP scale model ( <a href="https://www.ncbi.nlm.nih.gov/pmc/articles/PMC2676888/">https://www.ncbi.nlm.nih.gov/pmc/articles/PMC2676888/</a> )
27	HPLC_Retention_Ph_2_1	Total value of retention coefficients in HPLC at pH 2.1 (Proc. Natl. Acad. Sci. USA 77:1632-1636(1980))

28	HPLC_Tfa_Retention	Total value of retention coefficients in HPLC/TFA (Anal. Biochem. 124:201-208(1982))
29	Hplc_Hfba_Retention	Total value of retention coefficients in HFBA (Anal. Biochem. 124:201-208(1982))
30	Hplc_Retention_Ph_7_4	Total value of retention coefficients in HPLC at pH 7.4 (Proc. Natl. Acad. Sci. USA 77:1632-1636(1980))
31	Hydrophobicity_Abraham_Leo	Total hydrophobicity, as defined by Abraham and Leo (Proteins: Structure, Function and Genetics 2:130-152(1987))
32	Hydrophobicity_Black	Total hydrophobicity, as defined by Black (Anal. Biochem. 193:72-82(1991))
33	Hydrophobicity_Bull_Breese	Total hydrophobicity, as defined by Bull and Breese (Arch. Biochem. Biophys. 161:665-670(1974))
34	Hydrophobicity_Chothia	Total hydrophobicity based on proportion of buried residues (95%), as defined by Chothia (J. Mol. Biol. 105:1-14(1976))
35	Hydrophobicity_Eisenberg	Total normalized consensus hydrophobicity, as defined by Eisenberg et al. (J. Mol. Biol. 179:125-142(1984))
36	Hydrophobicity_Fauchere	Total hydrophobicity, as defined by Fauchere (Eur. J. Med. Chem. 18:369-375(1983))
37	Hydrophobicity_Guy	Total hydrophobicity based on free energy of transfer, as defined by Guy (Biophys J. 47:61-70(1985))
38	Hydrophobicity_Hopp_Woods	Total hydrophilicity, as defined by Hopp & Woods (Proc. Natl. Acad. Sci. U.S.A. 78:3824-3828(1981))
39	Hydrophobicity_Hplc_Parker	Total hydrophilicity derived from HPLC peptide retention times, as defined by Parker et al. (Biochemistry 25:5425-5431(1986))
40	Hydrophobicity_Hplc_Ph_3_4_Cowan	Total hydrophobicity determined by HPLC at pH 3.4, as defined by Cowan and Whittaker (Peptide Research 3:75-80(1990))
41	Hydrophobicity_Hplc_Ph_7_5_Cowan	Total hydrophobicity determined by HPLC at pH 7.5, as defined by Cowan and Whittaker (Peptide Research 3:75-80(1990))
42	Hydrophobicity_Hplc_Wilson	Total hydrophobicity derived from HPLC peptide retention times, as defined by Wilson et al. (Biochem. J. 199:31-41(1981))
43	Hydrophobicity_Janin	Total hydrophobicity based on dG of transfer from inside to outside of a globular protein, as defined by Janin (Nature 277:491-492(1979))
44	Hydrophobicity_Kyte_Doolittle	Total hydrophobicity, as defined by Kyte and Doolittle (J. Mol. Biol. 157:105-132(1982))
45	Hydrophobicity_Manavalan	Total average surrounding hydrophobicity, as defined by Manavalan and Ponnusamy (Nature 275:673-674(1978))
46	Hydrophobicity_Miyazawa_Jernigan	Hydrophobicity, as defined by Miyazawa and Jernigan (Macromolecules 18:534-552(1985))
47	Hydrophobicity_Rao_Argos	Total transmembrane helix parameters, as defined by Rao and Argos (Biochim. Biophys. Acta 869:197-214(1986))
48	Hydrophobicity_Rf_Mobility	Total hydrophobicity based on chromatographic mobility, as defined by Aboderin (Int. J. Biochem. 2:537-544(1971))
49	Hydrophobicity_Rose	Total hydrophobicity based on mean fractional exposed area loss (average area buried/standard state area), as defined by Rose (Science 229:834-838(1985))
50	Hydrophobicity_Roseman	Total hydrophobicity, as defined by Roseman (J. Mol. Biol. 200:513-522(1988))
51	Hydrophobicity_Sweet	Total optimized matching hydrophobicity, as defined by Sweet (J. Mol. Biol. 171:479-488(1983))
52	Hydrophobicity_Tanford	Total hydrophobicity, as defined by Tanford (J. Am. Chem. Soc. 84:4240-4274(1962))
53	Hydrophobicity_Welling	Total antigenicity, as defined by Welling (FEBS Lett. 188:215-218(1985))
54	Hydrophobicity_Wolfenden	Total hydration potential at 25 °C, as defined by Wolfenden (Biochemistry 20:849-855(1981))
55	Molecular_Weight	Molecular weight based on the sum of each amino acid molecular weight
56	Number_Of_Codons	Number of codons encoding each amino acid in the universal genetic code
57	Parallel_Beta_Strand	Parallel beta strand propensity, as defined by Lifson and Sander (Nature 282:109-111(1979))
58	Percentage_Accessible_Res	Total molar fraction of accessible residues, as defined by Janin (Nature 277:491-492(1979))

59	Percentage_Buried_Res	Total molar fraction of buried residues, as defined by Janin (Nature 277:491-492(1979))
60	Polarity_Grantham	Total polarity, as defined by Grantham (Science 185:862-864(1974))
61	Polarity_Zimmerman	Total polarity, as defined by Zimmerman (J. Theor. Biol. 21:170-201(1968))
62	Ratio_Hetero_End_Side	Total atomic weight ratio of hetero elements in end group to C in side chain (Science 185:862-864(1974))
63	Recognition_Factors	Total recognition factor of each amino acid, as defined by Fraga (Can. J. Chem. 60:2606-2610(1982))
64	Refractivity	Total refractivity index of each amino acid, as defined by Jones (J. Theor. Biol. 50:167-184(1975))
65	Relative_Mutability	Total relative mutability (Ala=100), as defined by Dayhoff et al. (In "Atlas of Protein Sequence and Structure", Vol.5, Suppl.3 (1978))
66	Theoretical pI*	Isoelectric point
67	Total_Beta_Strand	Total (antiparallel+parallel) beta strand propensity, as defined by Lifson and Sander (Nature 282:109-111(1979))
68	Transmembrane_Tendency	Total transmembrane tendency, as defined by Zhao and London (Protein Sci. 15:1987-2001(2006))

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161      **Table S2.** Summary of descriptor importance and model performance for a VH-based LogisticReg  
 162      model trained on all descriptors (excluding charge at pH 6 and 7.4). For each descriptor, the following  
 163      is provided; (i) cluster number assigned by hierarchical clustering of Spearman correlation  
 164      coefficients as a way to group redundant descriptors, (ii) LogisticReg coefficients from the logistic  
 165      regression model, (iii) the decrease in model accuracy when the descriptor is permuted, indicating its importance (based on  
 166      10-fold CV on test data), (iv) the model accuracy when the specific descriptor is left out, indicating  
 167      its unique contribution, and (v) the accuracy of the model using only the single descriptor. Descriptors  
 168      are listed along with their respective cluster number and their importance metrics, highlighting their  
 169      contribution to the model's performance and their individual predictive power.  
 170

#	Descriptor	Cluster number	LogisticReg coeff.	LogisticReg abs(coeff.)	Permutation on test data [%-units decrease in accuracy]	Leave-one-feature-out accuracy [%]	Single-descriptor model accuracy [%]
1	Disorder_Propensity_DisProt	7	0.51	0.51	2.9	70.0	50.2
2	Disorder_Propensity_TOP_IDP	12	0.46	0.46	2.5	69.8	48.8
3	Theoretical pI	22	0.45	0.45	-3.5	68.6	65.2
4	Aggrescan_av4	8	0.37	0.37	-0.2	70.0	55.4
5	Percentage_Accessible_Res	6	-0.36	0.36	1.4	70.1	52.0
6	Zygggregator_profile_smoothed_pos	14	0.35	0.35	-1.7	69.5	57.1
7	Hplc_Hfba_Retention	8	0.33	0.33	-4.3	70.1	57.3
8	Hydrophobicity_Tanford	8	-0.31	0.31	0.3	70.0	53.1
9	Hydrophobicity_Hplc_Ph_3_4_Cowan	8	-0.29	0.29	1.6	70.1	54.6
10	Polarity_Zimmerman	15	-0.27	0.27	0.0	70.0	61.2
11	Aggrescan_Nr_hotspots	17	-0.26	0.26	0.0	69.1	49.3
12	Hydrophobicity_Hplc_Ph_7_5_Cowan	8	0.25	0.25	0.8	70.0	57.6
13	Beta_Sheet_Chou_Fasman	10	0.24	0.24	-0.6	70.0	54.8
14	Hydrophobicity_Welling	18	-0.22	0.22	-0.6	69.7	54.7
15	Refractivity	21	-0.22	0.22	3.0	70.0	49.4
16	Ratio_Hetero_End_Side	19	-0.21	0.21	2.1	70.0	53.1
17	Average_Flexibility_BP	2	0.21	0.21	1.1	70.0	54.6
18	Beta_Turn_Chou_Fasman	4	-0.20	0.20	3.3	69.8	50.9
19	Hydrophobicity_Hplc_Parker	8	-0.20	0.20	0.0	70.0	57.0
20	Hydrophobicity_Roseman	8	0.19	0.19	2.1	70.0	57.1
21	Beta_Turn_Deleage_Roux	5	0.19	0.19	-0.6	70.0	55.3
22	Aa_Composition_Swissprot	9	0.19	0.19	0.6	70.0	52.9
23	Hydrophobicity_Eisenberg	8	-0.19	0.19	0.8	70.0	53.7
24	Number_Of_Codons	9	0.19	0.19	2.7	70.0	56.8
25	Bulkiness	10	0.17	0.17	-1.3	70.0	56.8
26	Hydrophobicity_Sweet	7	0.16	0.16	0.8	70.0	50.8
27	Alpha_Helix_Levitt	1	0.15	0.15	-0.3	70.0	49.8
28	Hydrophobicity_Janin	8	-0.13	0.13	-0.2	70.0	54.1
29	Beta_Turn_Levitt	4	-0.12	0.12	3.5	70.0	50.1
30	Aa_Composition	9	0.12	0.12	1.0	70.0	52.7
31	Hydrophobicity_Hopp_Woods	7	-0.12	0.12	1.9	70.0	55.6
32	Hydrophobicity_Fauchere	8	-0.12	0.12	1.9	70.0	54.7

33	Hydrophobicity_Rao_Argos	10	-0.12	0.12	1.9	70.0	53.1
34	Recognition_Factors	2	-0.11	0.11	3.2	70.0	53.2
35	Alpha_Helix_Chou_Fasman	1	-0.11	0.11	2.1	70.0	50.4
36	Hydrophobicity_Abraham_Leo	8	-0.11	0.11	2.7	70.0	52.4
37	Aggrescan_av4_pos	8	0.10	0.10	1.3	69.8	53.4
38	Relative_Mutability	20	0.09	0.09	-1.3	70.0	50.9
39	Hydrophobicity_Rf_Mobility	10	-0.09	0.09	2.7	70.0	54.4
40	HPLC_Tfa_Retention	8	0.08	0.08	-0.5	70.0	53.2
41	HPLC_Retention_Ph_2_1	7	-0.07	0.07	3.2	70.0	54.9
42	Hydrophobicity_Guy	8	-0.07	0.07	1.1	70.0	51.3
43	Aa_Flexibility_VTR	2	0.07	0.07	-0.5	70.0	51.7
44	Hydrophobicity_Manavalan	10	0.06	0.06	-1.0	70.0	52.8
45	Hydrophobicity_Chothia	10	-0.06	0.06	2.4	70.0	55.1
46	Hydrophobicity_Bull_Breese	7	0.06	0.06	1.3	70.1	50.6
47	Hydrophobicity_Miyazawa_Jernigan	10	0.06	0.06	-0.5	70.0	53.3
48	Alpha_Helix_Deleage_Roux	1	-0.05	0.05	2.2	70.0	53.1
49	Beta_Sheet_Levitt	10	-0.05	0.05	3.7	70.0	54.8
50	Hydrophobicity_Black	8	-0.04	0.04	3.0	70.0	53.9
51	Disorder_Propensity_FoldUnfold	10	-0.04	0.04	3.2	70.0	49.7
52	Hydrophobicity_Kyte_Doolittle	8	-0.03	0.03	0.2	70.0	53.2
53	Percentage_Buried_Res	10	-0.03	0.03	0.6	70.0	51.2
54	Hydrophobicity_Hplc_Wilson	13	-0.03	0.03	3.0	70.0	47.6
55	Polarity_Grantham	23	0.02	0.02	-1.4	70.0	52.0
56	Total_Beta_Strand	10	-0.02	0.02	2.9	70.0	53.7
57	Zyggregator_profile_smoothed	14	-0.02	0.02	1.7	70.1	57.3
58	Parallel_Beta_Strand	10	-0.02	0.02	2.9	70.0	54.2
59	Hydrophobicity_Rose	10	-0.02	0.02	2.7	70.0	51.0
60	Antiparallel_Beta_Strand	10	-0.02	0.02	2.7	70.0	53.4
61	Hydrophobicity_Wolfenden	16	0.01	0.01	1.0	70.0	52.9
62	Coil_Deleage_Roux	3	-0.01	0.01	1.6	70.0	52.8
63	Hplc_Retention_Ph_7_4	7	0.01	0.01	0.6	70.0	56.5
64	Beta_Sheet_Deleage_Roux	10	0.01	0.01	0.3	70.0	55.1
65	Transmembrane_Tendency	8	0.00	0.00	0.0	70.0	54.9
66	Avg_Area_Buried	11	0.00	0.00	0.0	70.0	50.1