

2019 Spring @ <https://qdata.github.io/deep2Read/>

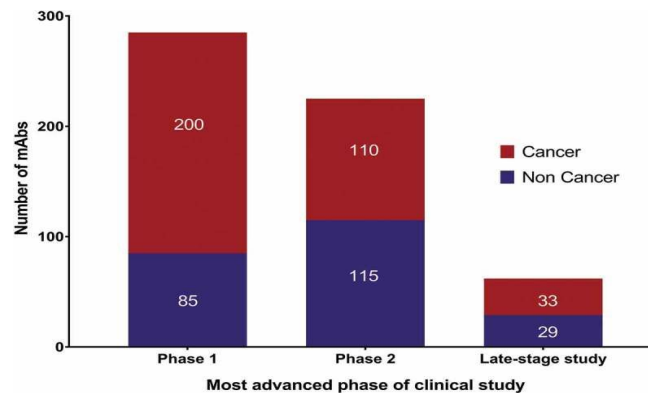
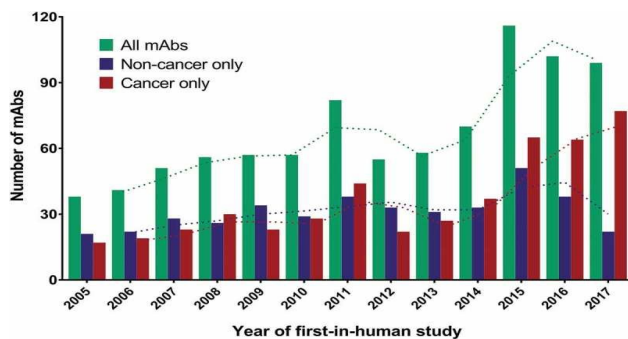
Attentive cross-modal paratope prediction

Andreea Deac, Petar Velickovic, Pietro Sormanni

Presented By Eli Draizen

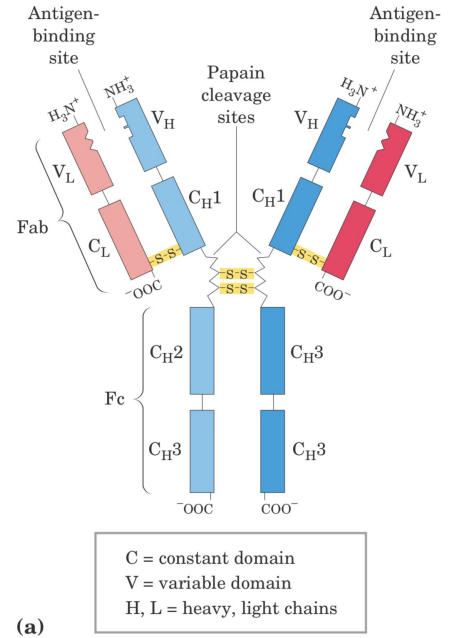
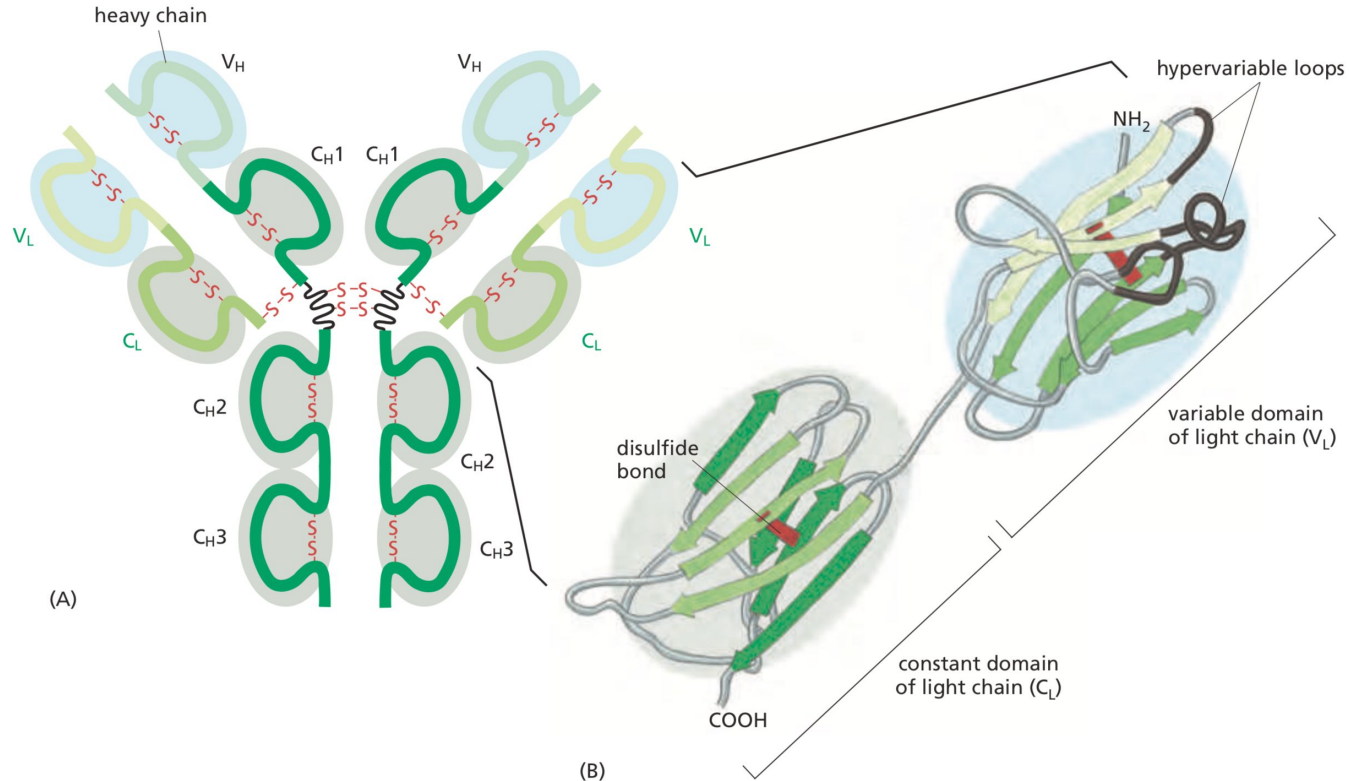
Background

- Antibodies are easier to create than small molecules
 - Created via phage display, ‘evolution in the lab’
 - More variable
- Antibodies as drugs are becoming more popular
- The challenge becomes creating binding site sequences in Antibody

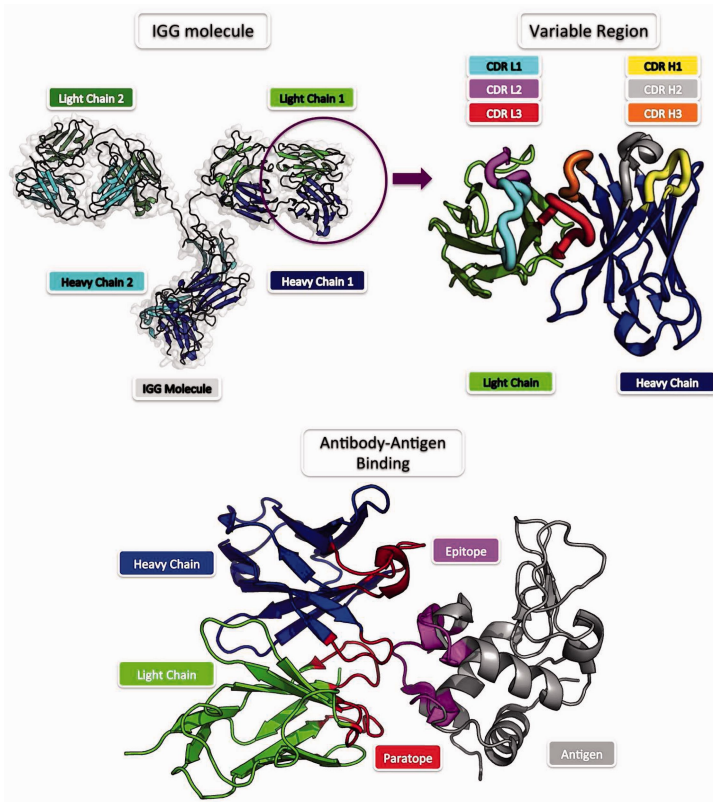


As of November 2018

Antibody structure



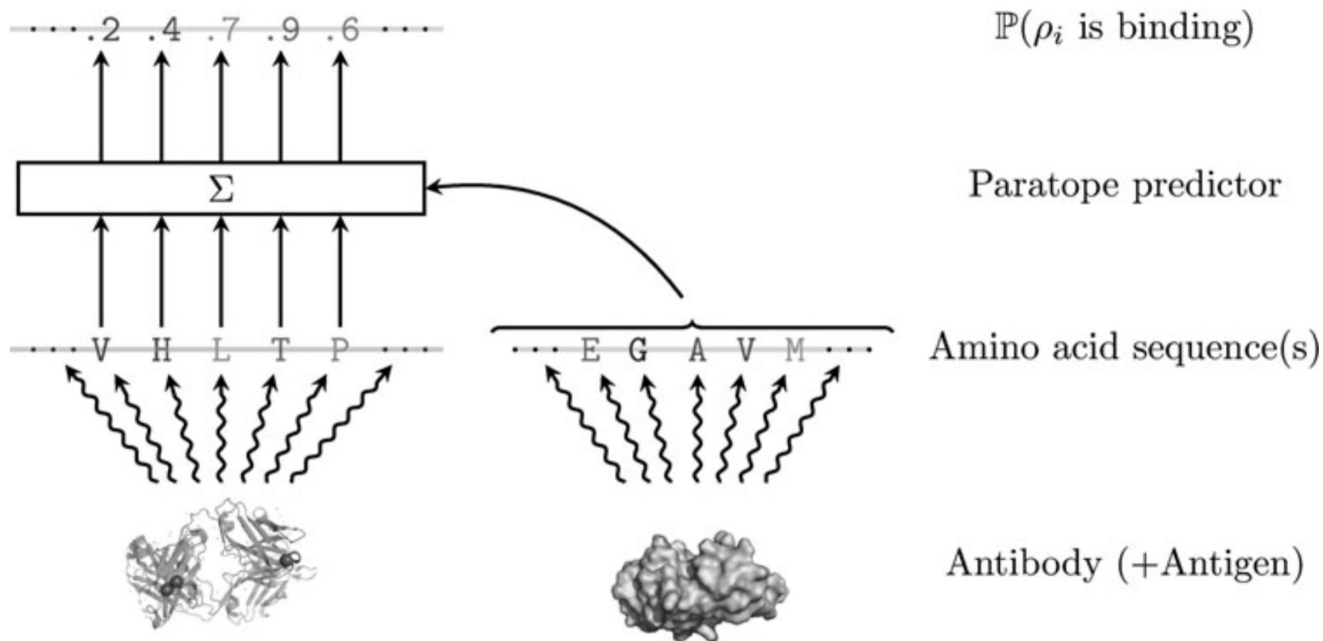
Definitions



- **Paratope:** Binding site on antibody
 - On average 40–50 residues, while only around 18–19 residues are in contact with antigen
- **Epitope:** Binding site on antigen
- **Antigen:** Target protein (E.g. from parasite)

Problem Statement

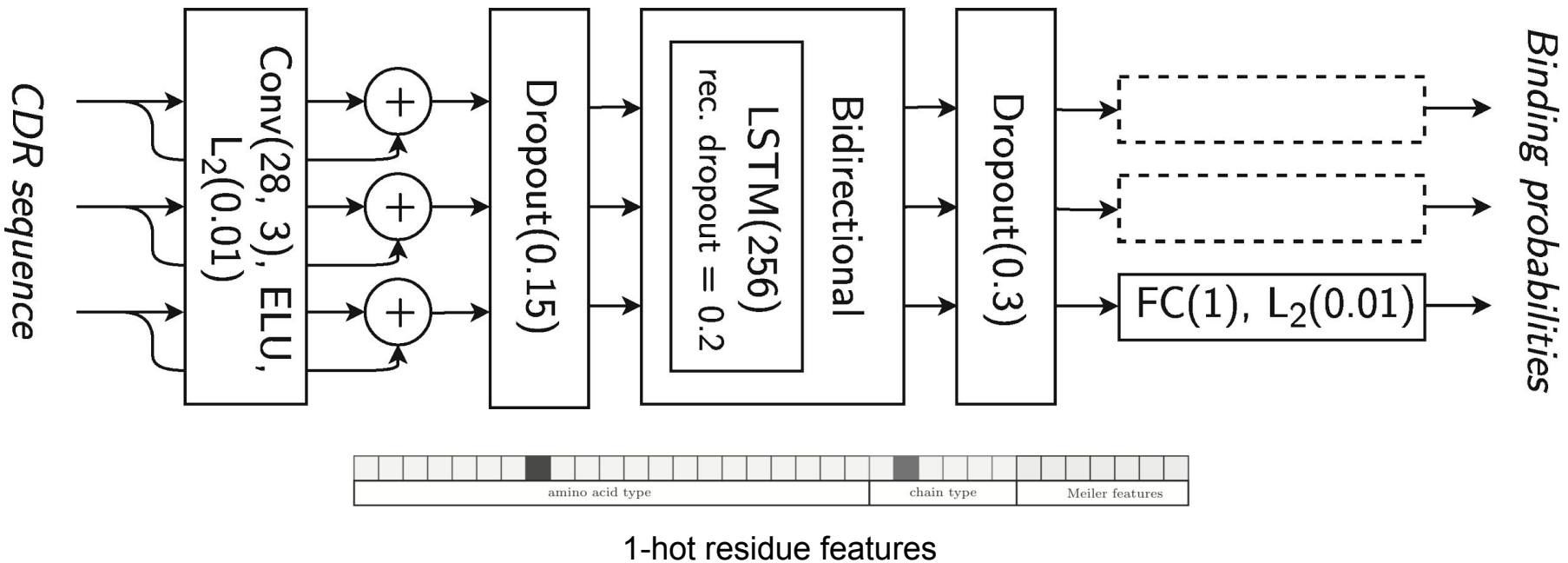
Goal: Given a CDR and antigen sequence, predict if it will bind an antigen



Previous Approaches

- Template-based: sequence or structural homology from alignments of to one side of interface with Ab-Ab structures
- Residue propensities of known Ab-Ag complexes
 - Probability distributions based on a Kolmogorov-Smirnov test (Ab-i-Patch)

Parapred



Fast-ParaPred

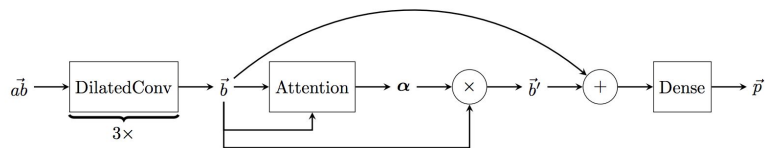
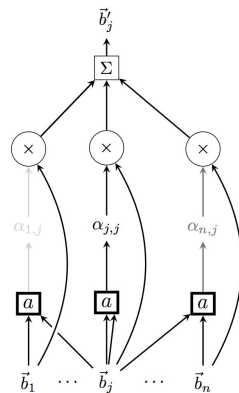
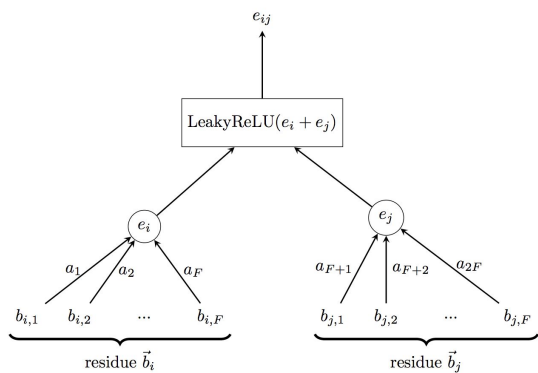


Figure 1. The Fast-ParaPred architecture.



- Substitute RNN with:
 - Dilated Convolutions
 - 3 layers 64,128,256 feats with rates 1,2,4
 - Self-attention layer
- **ab** is AntiBody CDR sequence
- **p** is classification of binding site or not

AG-Fast-Parapred

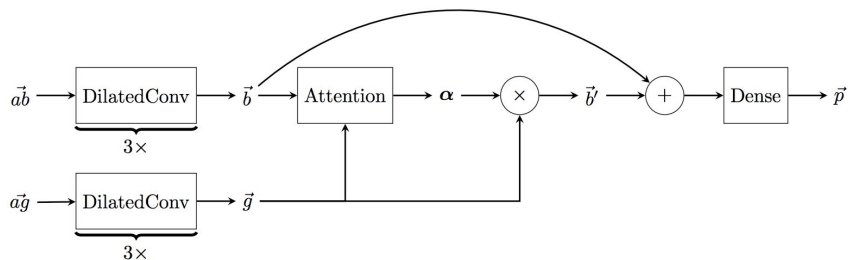
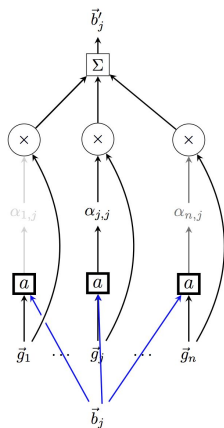


Figure 4. The AG-Fast-Parapred architecture.



- Include all residues from **AntiGen**
- Same Dilated Convolutions as with **ab**
- Update attention weights where **ab** is the query and **ag** as key/value:

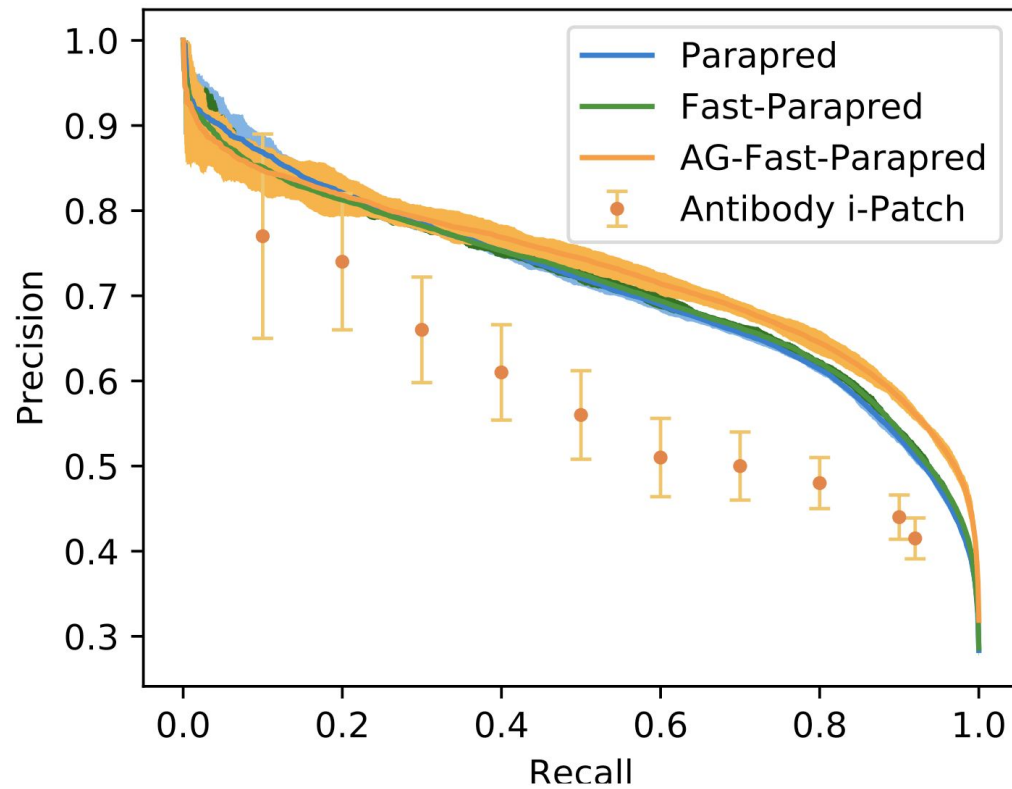
$$\alpha_{ij} = \frac{\exp\left(\text{LeakyReLU}\left(\vec{\mathbf{a}}^T[\mathbf{W}_1\vec{b}_i \parallel \mathbf{W}_2\vec{g}_j]\right)\right)}{\sum_{k \in \nu_i} \exp\left(\text{LeakyReLU}\left(\vec{\mathbf{a}}^T[\mathbf{W}_1\vec{b}_i \parallel \mathbf{W}_2\vec{g}_k]\right)\right)} \quad (5)$$

Experiments

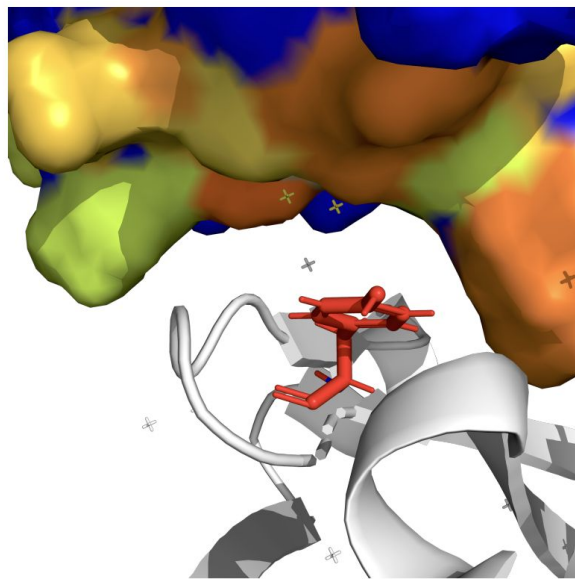
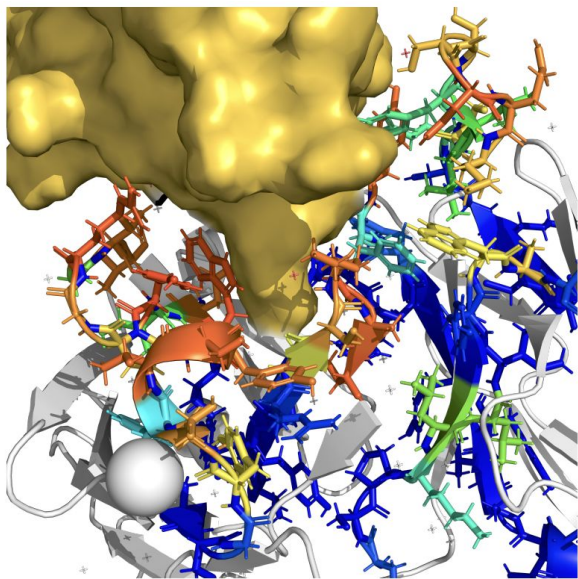
- Data collected from Structural Antibody Database (SAbDB)
 - All: 3522 pdb structures
 - Filtered: ~304 structures
- Results after 10 rounds of 10-fold Cross Validation:

Method	ROC AUC	MCC	Epoch time
proABC (Olimpieri et al., 2013)	0.851	0.522	—
Parapred (Liberis et al., 2018)	0.880 ± 0.002	0.564 ± 0.007	$0.190 \pm 0.019s$
Fast-Parapred (ours)	0.883 ± 0.001	0.572 ± 0.004	$0.085 \pm 0.015s$
AG-Fast-Parapred (ours)	0.899 ± 0.004	0.598 ± 0.012	$0.178 \pm 0.020s$

Results



Results



Left: Ab-Ag complex.

Right: Normalized Ag attention weights

Warmer colors indicate higher probabilities and coefficients

Conclusion

- Fast-Parapred outperforms current state of the art method and does not use any antigen sequence
- AG-Fast-Parapred outperforms Fast-Parapred since it uses knowledge of antigen sequences, which maybe at the binding site.
- No 3D coordinates used

Drawbacks:

- A CDR sequence is needed, not a generative model
- Trained on a small dataset

Relation to our project

1. Predict binding sites on target protein and potential superfamilies it may bind to using GraphNN
2. Only include those residues in AG-Fast-ParaPred

or

3. Find partner bind site with complimentary shape and charge
 - a. Search through examples of target superfamily with interacting superfamily
4. Given structure of antibody without variable regions, design new variable region with similar geometry and charge to the partner binding site or
 - a. Denoising Autoencoder/GAN