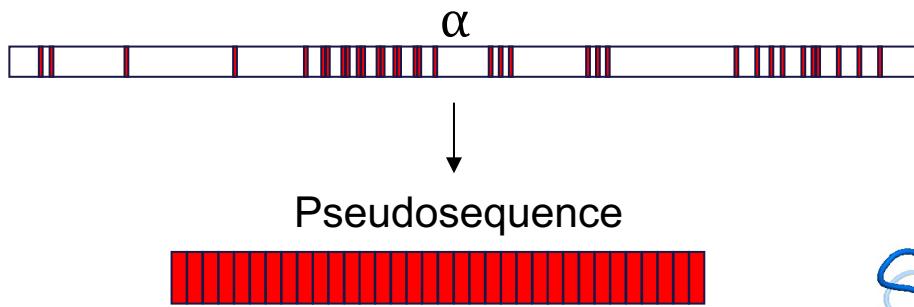


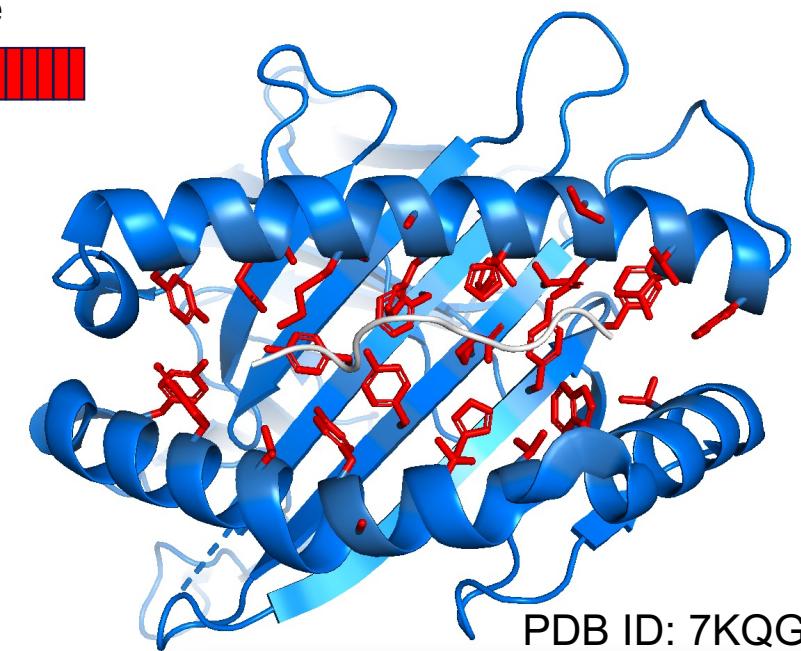
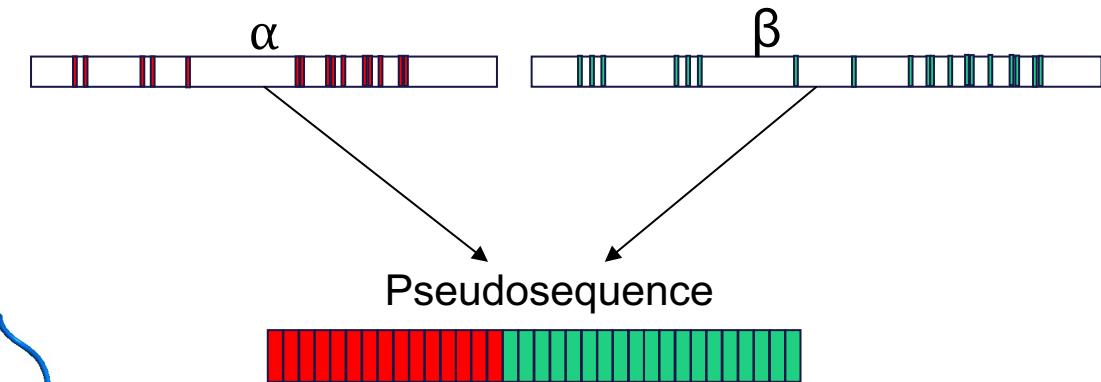
Stuff from my Master's thesis

MHC sequences

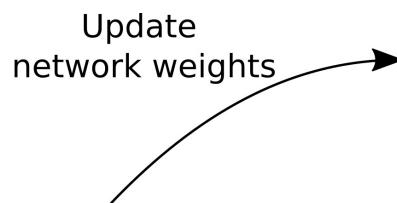
Class I: Alpha chain



Class II: alpha + beta chain



PDB ID: 7KQG



NNAlign-2.1

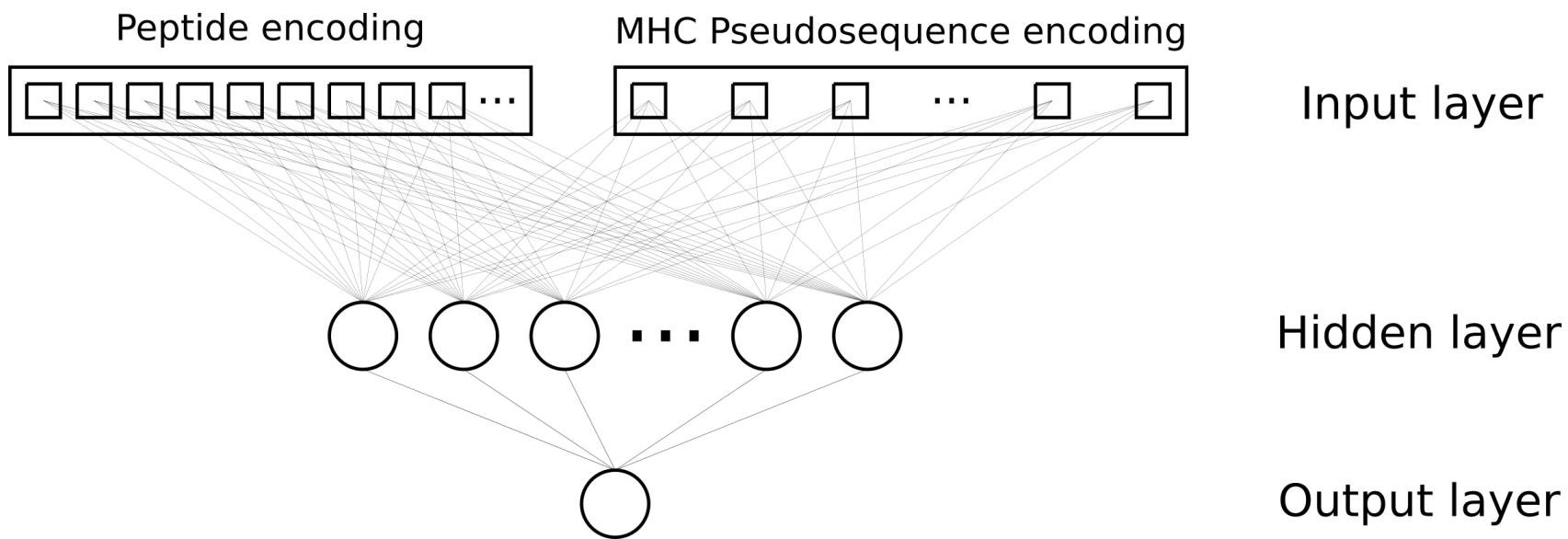
Predict binding cores
and affinities

Peptide	Measurement	Prediction	Error
DSY			
I P Q S L W T S L	0.679	0.619	0.060
- A L P P V A P V	0.503	0.507	0.002
F S L P F P F L Y K F L L	0.590	0.482	0.108
E V V M A Y V G I K	0.577	0.448	0.129
Q L L G K L P H L R M	0.010	0.020	0.010
- R G Y V F Q G L	0.198	0.206	0.008
R L R D L L L I V T R	0.059	0.023	0.036
⋮			

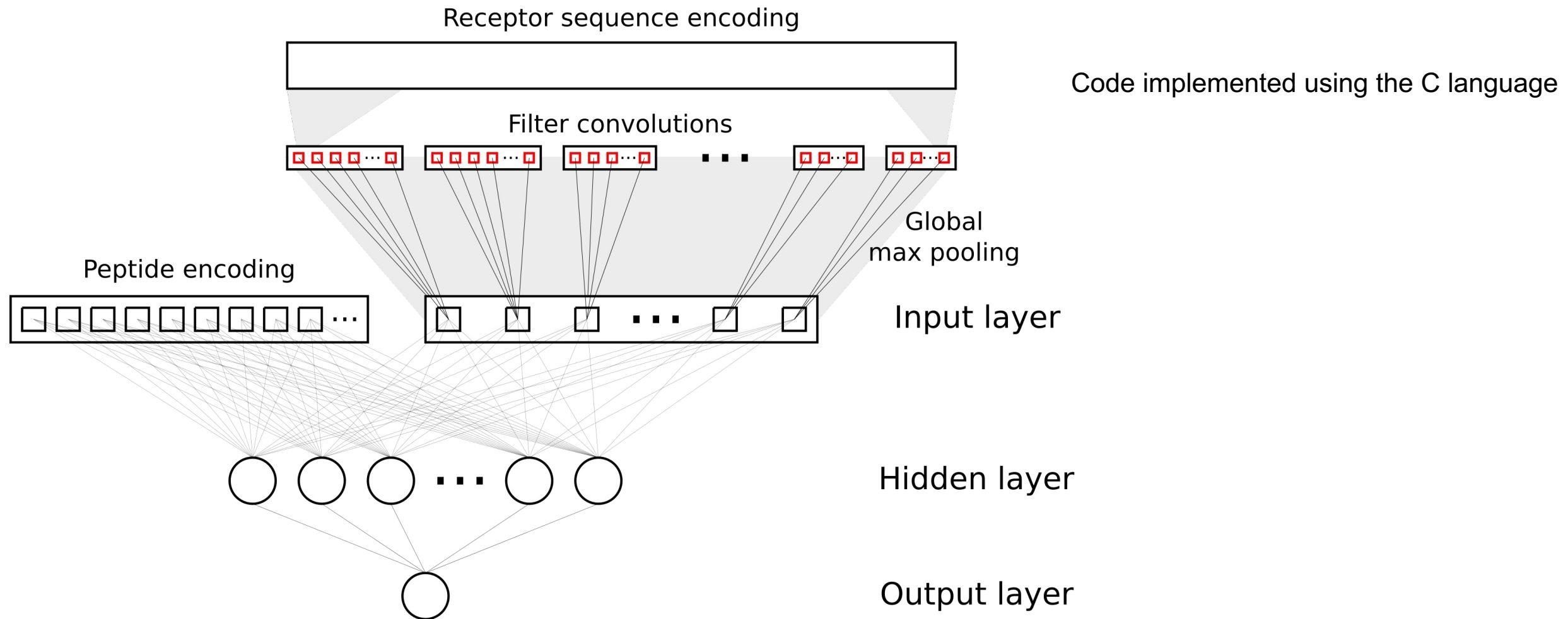
Calculate
prediction errors

Peptide	Measurement
ALPPVAPV	0.503
Binding core	
-ALPPVAPV	0.362
A-LPPVAPV	0.473
AL-PPVAPV	0.507
ALP-PVAPV	0.416
⋮	⋮
Peptide	Measurement
IPQSLDSYWTSL	0.679
Alignment	
IPQSLDSYW TSL	IPQSLDSYW
PQSLDSYWT SL	PQSLDSYWT
IP QSLDSYWT SL	QSLDSYWT
⋮	⋮
PQSLDSYWT SL	IQLSDSYWT
PQ SLDSYWT SL	IPSLD SYWT
IPQS LDSYWT SL	IPQLDSYWT
⋮	⋮
IPQS LDSYWT SL	IPQSYWTSL
IPQSL DSYWT SL	IPQSLWTSL
IPQSLD SYWT SL	IPQSLDTSL
⋮	⋮

The conventional NNAlign method

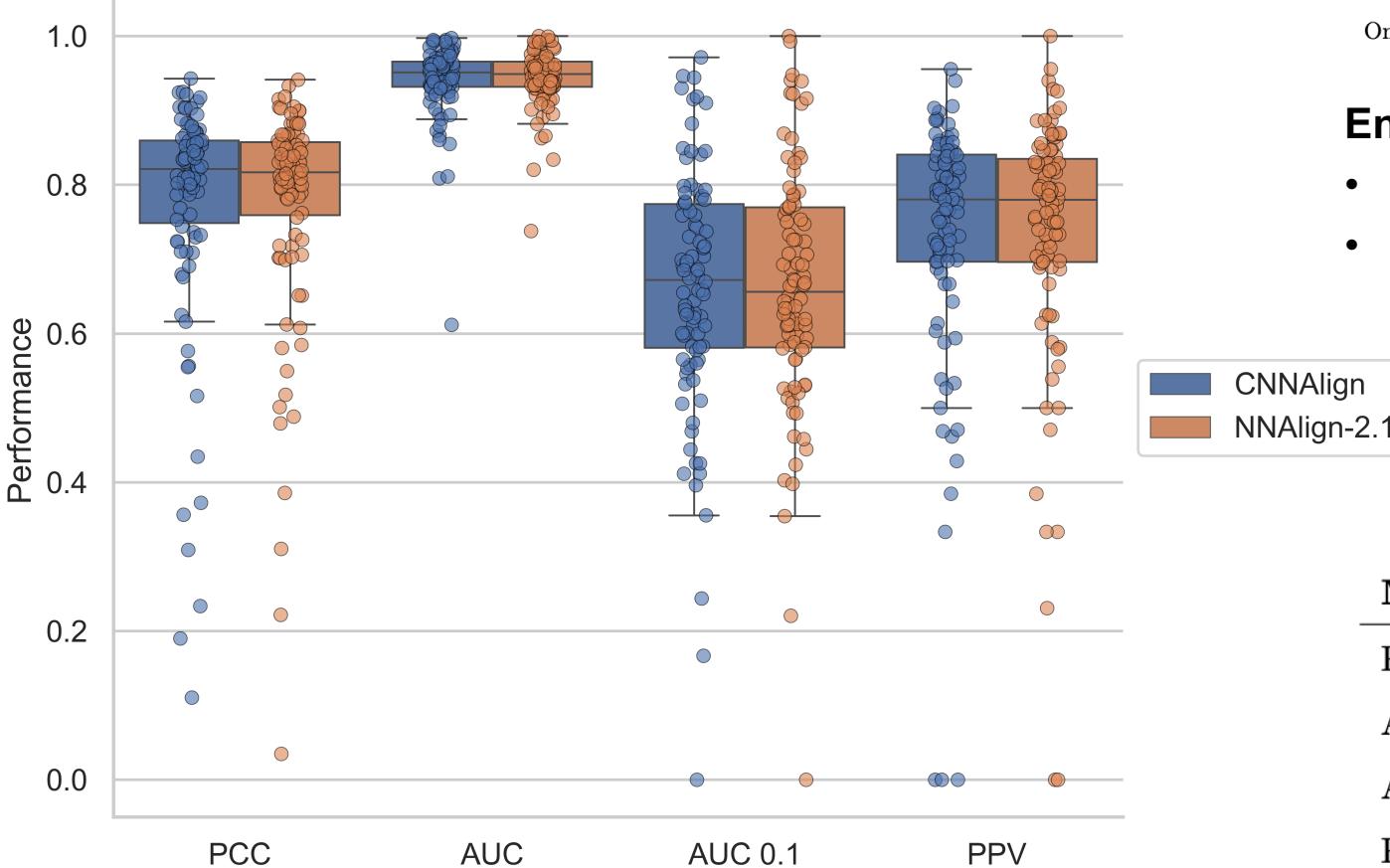


CNNAlign



Results: MHC class I

Comparison of per-allele performance metrics



Parameter	Value	Number of input values
Peptide encoding	Blosum	-
Motif (binding core) length	9	180 (9×20)
PFR length	0	-
PFR length encoding	False	-
Max length of insertions	1	-
Max length of deletions	10	-
Individual peptide lengths encoded	8,9,10,11,12,13,14	7
Encode insertion length	True	2
Encode deletion length	True	2
Only make insertions in peptides shorter than the motif length	True	-

Ensemble of 20 networks

- 5-fold CV
- 4 initializations per fold

CNN filters used (180 in total)

- 50 of length 10
- 50 of length 8
- 40 of length 5
- 40 of length 3

Metric	#CNNAlign wins	#Ties	P-value (without ties)
PCC	71	0	0.00001
AUC	60	2	0.00103
AUC 0.1	62	0	0.00056
PPV	38	21	0.31765

Positional importance

Rescaled filter weights (5 x 20)

	i=1	i=2	i=3	i=4	i=5
A	-0.55	0.51	0.23	-0.19	0.39
R	-0.10	0.02	2.11	-0.66	0.23
N	-0.59	-0.30	0.67	-0.91	1.02
D	-0.50	-0.14	0.67	-0.82	1.24
C	-0.58	0.70	0.34	0.35	0.45
Q	-0.49	-0.06	0.83	-0.77	0.46
E	-0.33	0.01	0.83	-0.73	0.92
G	-0.65	-0.01	0.71	-0.49	0.46
H	0.11	-0.39	0.77	-0.10	0.38
I	0.67	2.11	0.30	0.95	-0.05
L	0.81	1.78	0.59	1.18	-0.28
K	-0.41	-0.19	1.58	-0.97	0.53
M	0.44	1.48	0.33	0.80	-0.20
F	1.62	1.02	0.44	1.90	-0.57
P	-0.29	0.21	0.50	-0.25	0.78
S	-0.68	-0.08	0.50	-0.52	0.55
T	-0.38	0.58	0.33	-0.10	0.22
W	1.56	0.51	1.05	1.71	-0.12
Y	1.25	0.42	0.30	1.29	-0.47
V	0.41	1.97	0.27	0.74	0.02

$$\sigma_i = \text{std.dev}(s_{i,A}, s_{i,R}, \dots, s_{i,V})$$

$$\begin{array}{ccccc} \sigma_1 & \sigma_2 & \sigma_3 & \sigma_4 & \sigma_5 \\ 0.74 & 0.76 & 0.46 & 0.90 & 0.48 \end{array}$$

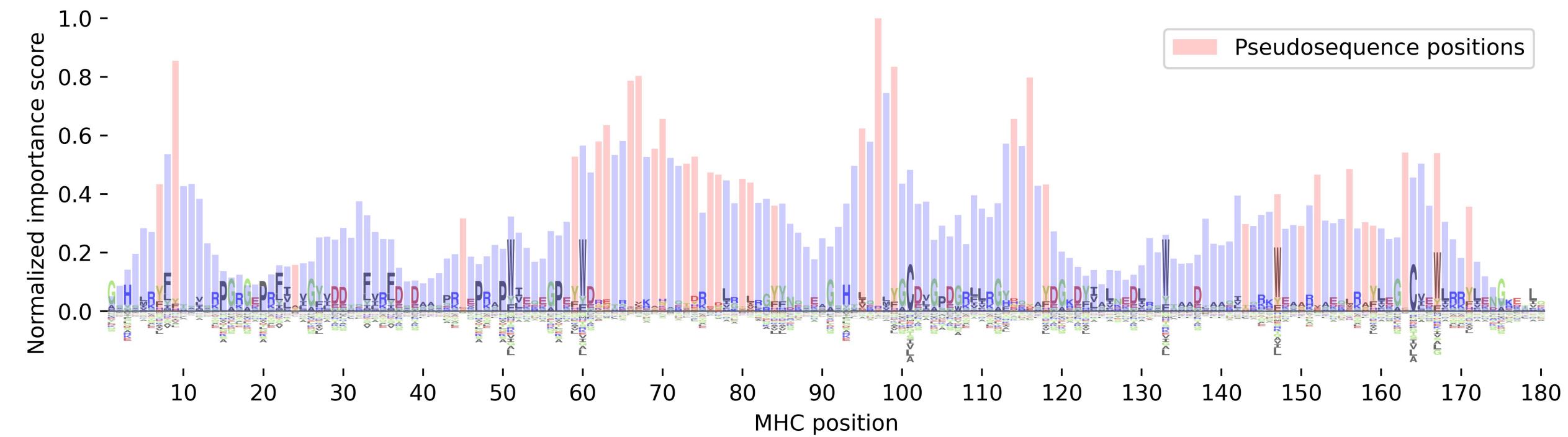
- Weight each position in the MHC based on how many filters had their max score around that position
- The weight depends on how much each filter has ‘learned’

$$\begin{array}{ccccccc} & 33 & 34 & 35 & 36 & 37 & \\ \dots & 0.00 & 0.00 & 0.00 & 0.00 & 0.00 & \dots \end{array}$$

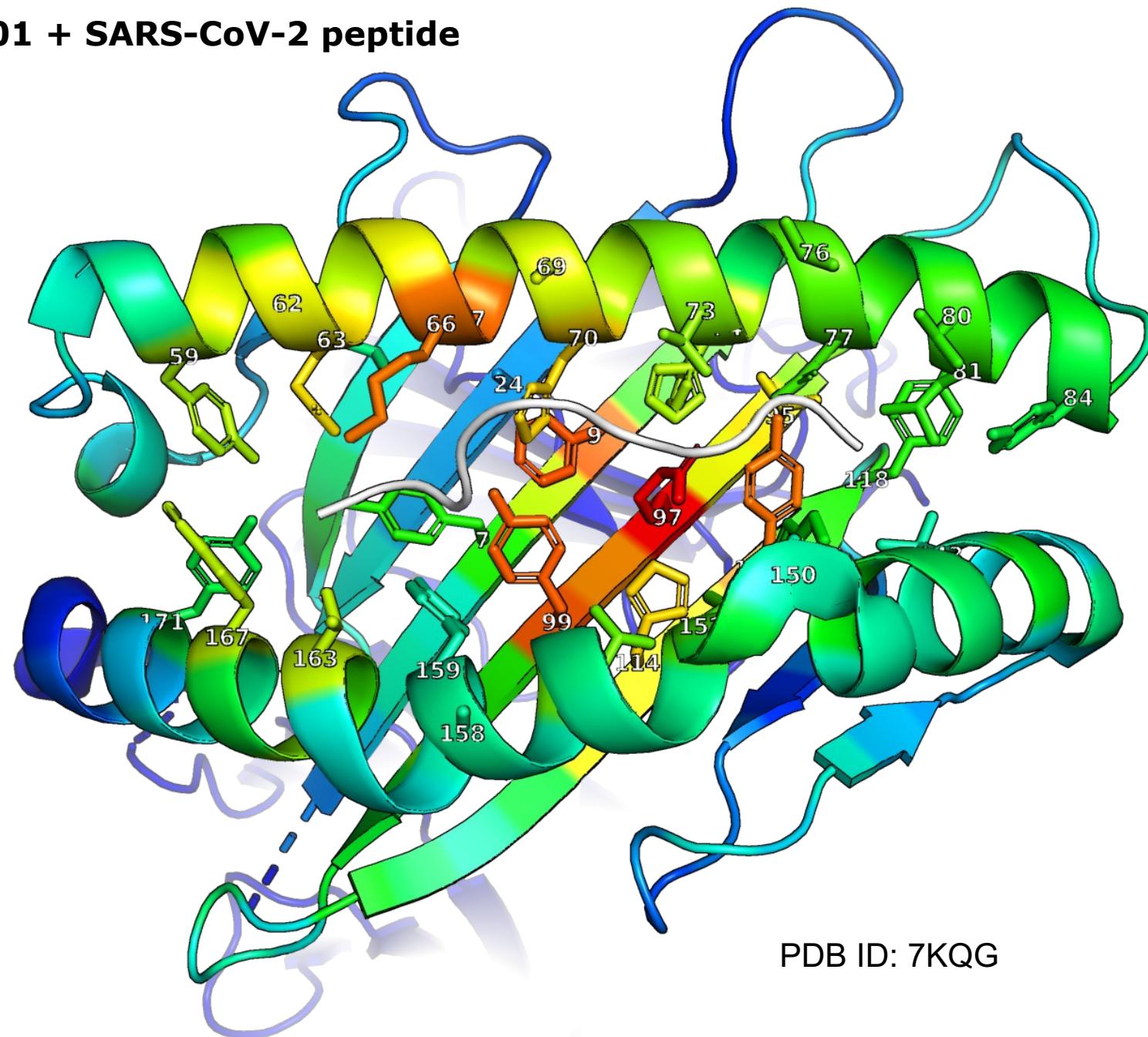
+

$$\begin{array}{ccccccc} & 0.19 & 0.19 & 0.12 & 0.23 & 0.12 & \\ \hline \end{array}$$

$$\begin{array}{ccccccc} & 33 & 34 & 35 & 36 & 37 & \\ \dots & 0.19 & 0.19 & 0.12 & 0.23 & 0.12 & \dots \end{array}$$

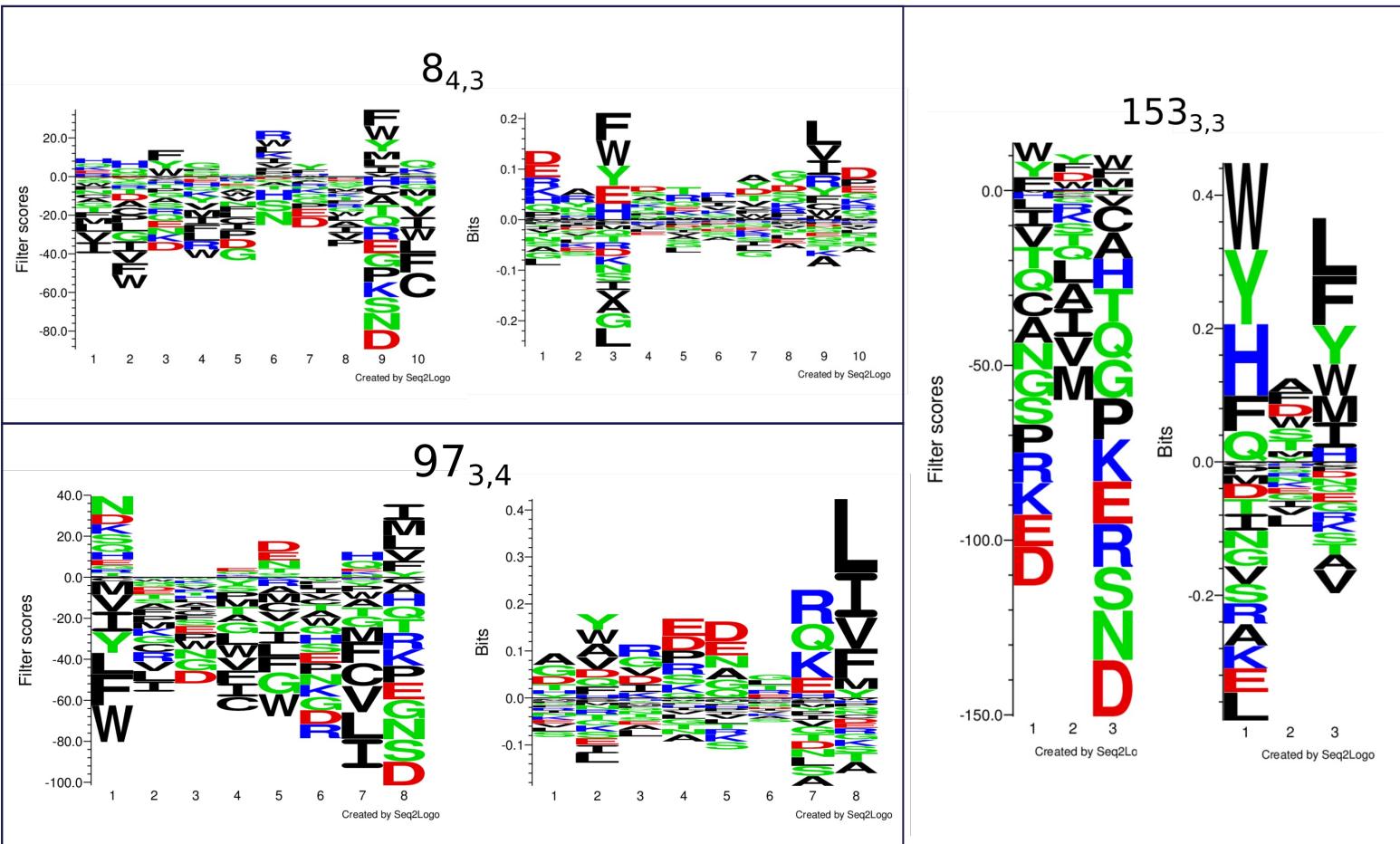


Sticks:
pseudosequence
positions



Filter sequence logos

Filters with high ‘knowledge’



Filters with low ‘knowledge’

