The Role of Hydrophobicity in Peptide-MHC Binding

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Overview

- Background on MHC Class I
- NetMHC-4.0 and NetMHCpan-4.1
- Our data analysis and flowcharts
- Hydrophobicity scale and calculations
- Results
- Conclusion



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MHC Class I proteins

- Immune surveillance of cells detect foreign material
- Cell surface protein that presents peptides
- Peptide is held in a groove, with ends in binding pockets
- Mechanics of binding depend on MHC variant
- Highly polymorphic

Applications - vaccine design, cancer neoantigen, viral disease severity

peptide N-term. domain a, heli peptide C-term.

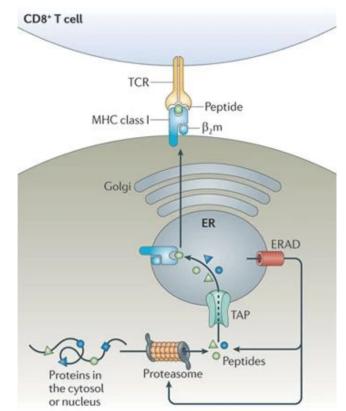
B*4402 (PDB:1M6O) Green, B*4405 (PDB: 1SYV) Blue



Antigen presentation pathway

- Endogenous protein
- Broken into small peptides
- Peptide binds to MHC class I protein (pMHC)
- If stable, transported to exterior
- Killer T cell binds to pMHC
 - If it is a foreign antigen -> infected cell is killed by T cell that recognizes it

As pMHC binding is the most selective step, it is the most studied process in determining cellular immune response.



Nature Reviews | Immunology

Neural Networks Prediction Tools

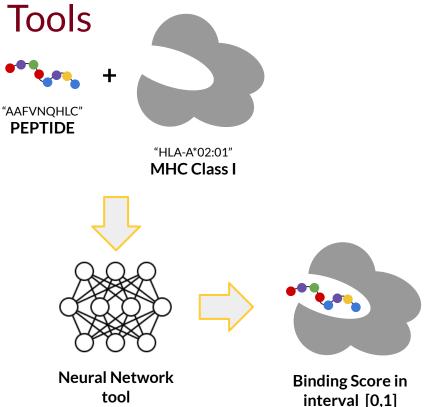
State-of-the-art tools:

- NetMHC-4.0
- NetMHCpan-4.1

Differences - MHC sequence, training data

Binding score in training data:

- Binding Affinity continuous
- Eluted Likelihood discrete



Massimo Andreatta, Morten Nielsen, Gapped sequence alignment using artificial neural networks: application to the MHC class I system, Bioinformatics, Volume 32, Issue 4, 15 February 2016, Pages 511–517, https://doi.org/10.1093/bioinformatics/btv639

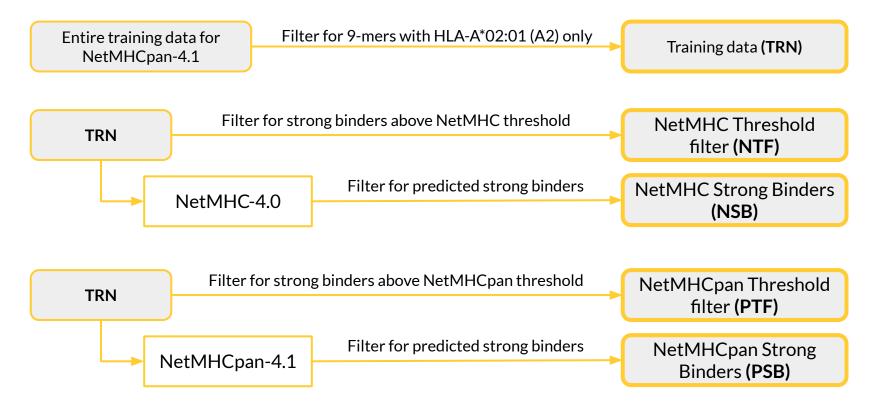
Birkir Reynisson, Bruno Alvarez, Sinu Paul, Bjoern Peters, Morten Nielsen, NetMHCpan-4.1 and NetMHCIIpan-4.0: improved predictions of MHC antigen presentation by concurrent motif deconvolution and integration of MS MHC eluted ligand data, Nucleic Acids Research, Volume 48, Issue W1, 02 July 2020, Pages W449–W454, https://doi.org/10.1093/nar/gkaa379

Two different data analyses

- Training data
- Human Proteome data

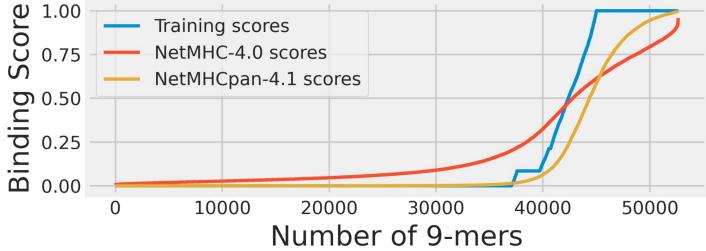
We only analyzed 9-mers, which show the strongest affinity for MHC Class I, and HLA-A*02:01 (A2), the most frequent MHC allele supertype

Training Data flowchart

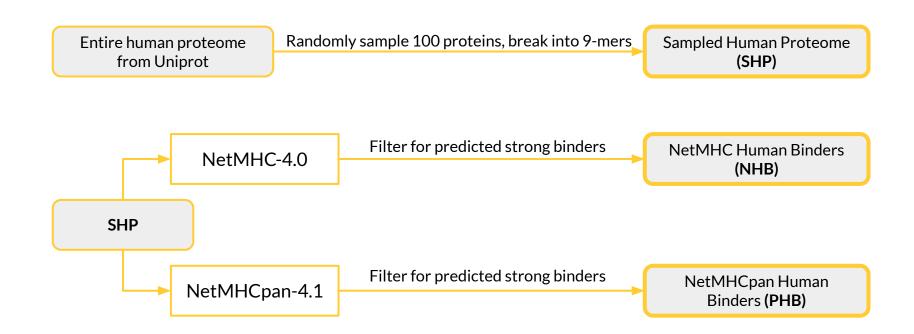


Training Data Scores

Sorted Scores for all A2 9-mers

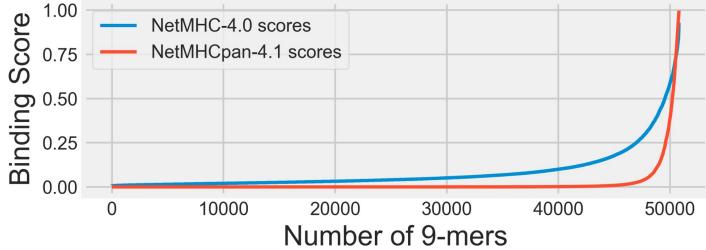


Human Proteome Data flowchart



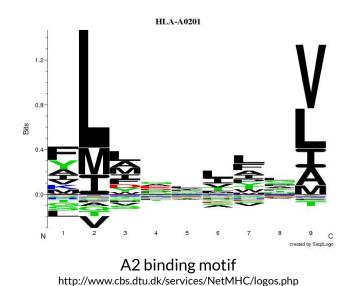
Human Proteome Scores

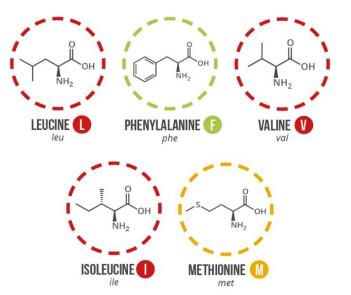
Sorted Scores for sampled human 9-mers



Delving into Hydrophobicity

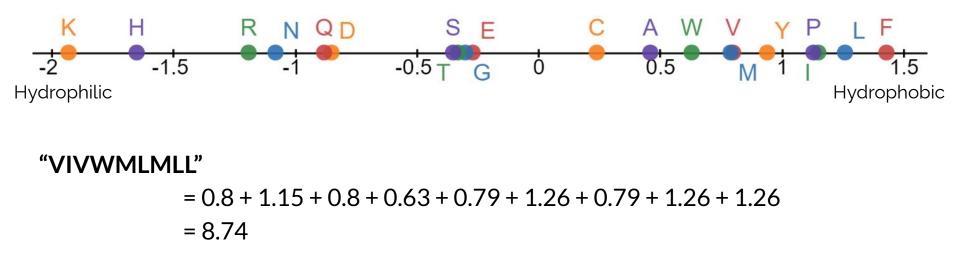
- Do either neural network tool utilize biochemical attributes in their predictions?
- The one we are most interested in is hydrophobicity. Why?





Moon Hydrophobicity Scale

- Models the polarity of the sidechain for an amino acid
- Hydrophobicity of peptide = sum of scores for each amino acid in it



Moon, C. Preston, and Karen G. Fleming. "Side-chain hydrophobicity scale derived from transmembrane protein folding into lipid bilayers." Proceedings of the National Academy of Sciences 108.25 (2011): 10174-10177.

Analysis of hydrophobicity

- For each set of peptides, calculate each peptide's hydrophobicity
- Statistical analysis across sets -
 - Graphical
 - Mean + STD
 - 2-sample T test

TRN: Training data

NTF & PTF: Strong binders in training data according to NetMHC and NetMHCpan

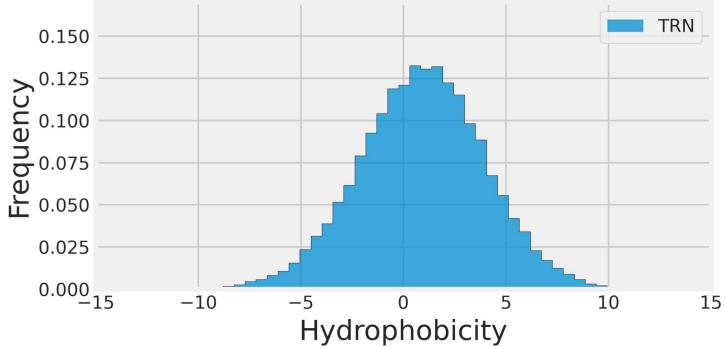
NSB & PSB: Predicted strong binders according to NetMHC and NetMHCpan

SHP: Sample Human Proteome data

NHB & PHB: Predicted strong binders according to NetMHC and NetMHCpan

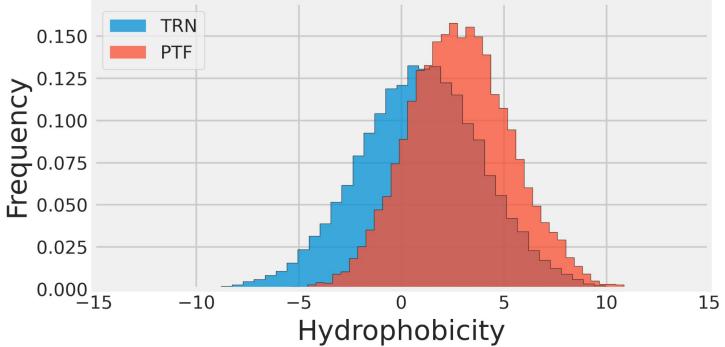
Results - NetMHCpan on training data

Histogram of Hydrophobicity of 9-mers



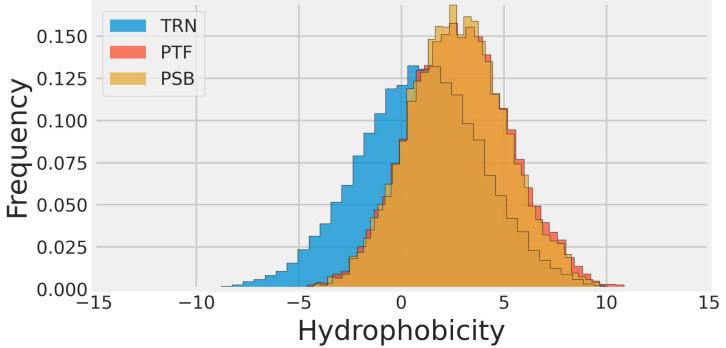
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Histogram of Hydrophobicity of 9-mers



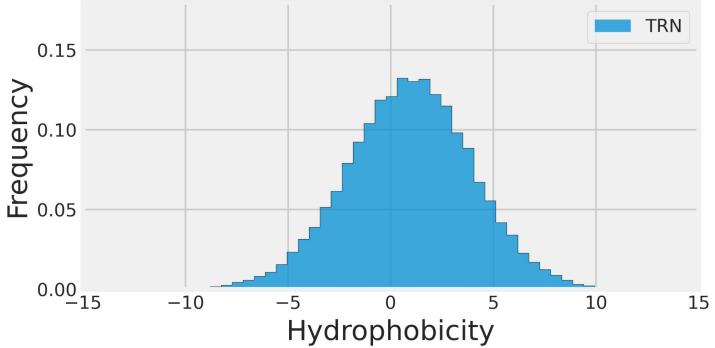
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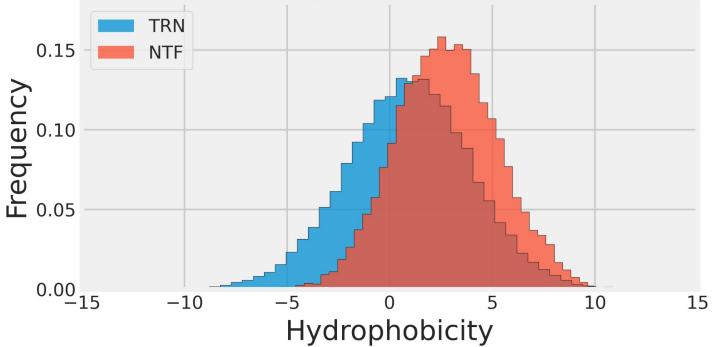
Results - NetMHC on training data

Histogram of Hydrophobicity of 9-mers



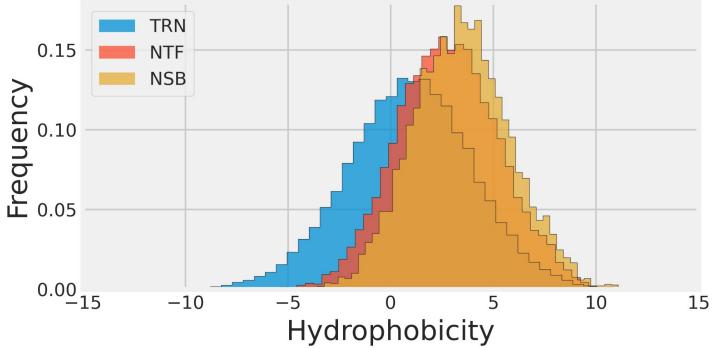
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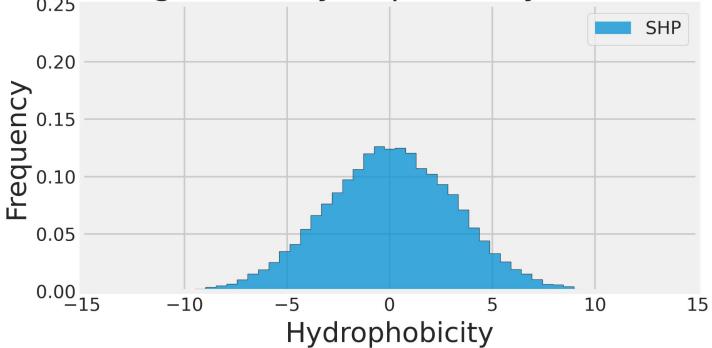
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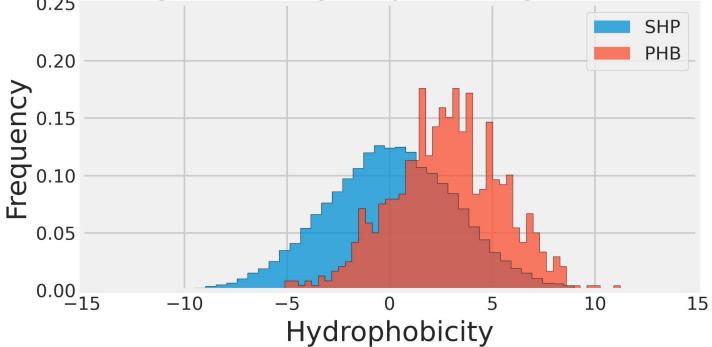
Results - Human Proteome

Histogram of Hydrophobicity of 9-mers

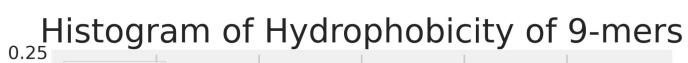


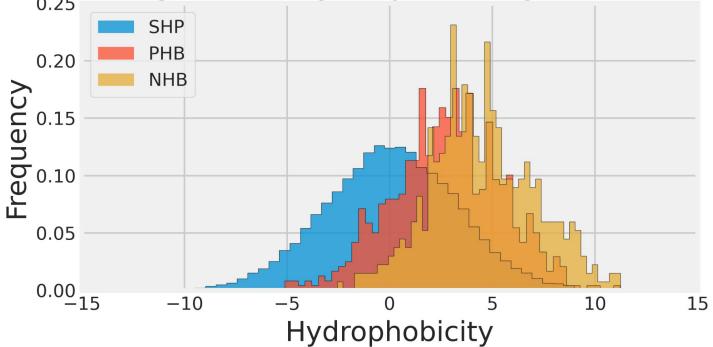
Results - Human Proteome

Histogram of Hydrophobicity of 9-mers



Results - Human Proteome





Conclusion

- NetMHC predicts strong binders to A2 to be more hydrophobic
 - Average of 3.45 vs 2.75 for NSB vs PSB
 - Average of 4.51 vs 2.78 for NHB vs PHB
- This bias is not reflected in
 - Training data
 - Prediction by NetMHCpan on training data
 - Prediction by NetMHCpan on the sample human proteome
 - Confirmed by statistical analysis for p value threshold of 0.0001
- The training data for NetMHCpan is not neutral in hydrophobicity shift by 1

Conclusion

- Neural Network tools are useful tools for pMHC prediction, but vulnerable to false positives
- Neural Network tools do not utilize biochemical attributes such as hydrophobicity in their calculation
- Improvements -
 - Hydrophobicity as a data feature
 - Better training data
 - Post processing filter