

Prediction of MHC Class Antigen Peptides from *Echinococcus Multilocularis*: Application of Computer Intelligence

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Abstract

Echinococcus multilocularis is a cyclophyllid cestode that causes hydatid disease in many mammals, including rodents and humans and is becoming an increasing problem in urban areas. Peptide fragments of antigen protein can be used to select nonamers for use in rational vaccine design and to increase the understanding of roles of the immune system in infectious diseases. Analysis shows MHC class II binding peptides of antigen protein from *Echinococcus multilocularis* are important determinant for protection of host from parasitic infection. In this assay, we used PSSM and SVM algorithms for antigen design and predicted the binding affinity of antigen protein having 426 amino acids, which shows 417 nonamers. Binding ability prediction of antigen peptides to major histocompatibility complex (MHC) class I & II molecules is important in vaccine development from *Echinococcus multilocularis*.

Keywords: Antigen protein; Epitope; PSSM; SVM; MHC; Peptide vaccine

Abbreviations: GES: Goldman, Engelberg and Steitz; MHC: major histocompatibility complex; PSSMs: Position Specific Scoring Matrices; SVM: Support Vector Machine

Introduction

Echinococcus multilocularis is a cyclophyllid cestode that causes hydatid disease in many mammals, including rodents and humans and is becoming an increasing problem in urban areas [1,2]. *Echinococcus multilocularis* antigen peptides are most suitable for subunit vaccine development because with single epitope, the immune response can be generated in large population. This approach is based on the phenomenon of cross-protection, whereby a plant infected with a mild strain of virus is protected against a more severe strain of the same virus. The phenotype of the resistant transgenic hosts includes fewer centers of initial virus infection, a delay in symptom development, and low virus accumulation. Antigen protein from *Echinococcus multilocularis* is necessary for new paradigm of synthetic vaccine development and target validation [3-5].

Methodology

In this research work antigenic epitopes of antigen protein from *Echinococcus multilocularis* is determined using the Gomase in 2007, Bull & Breeze, Eisenberg, Rao & ArgosChou & Fasman and Levitt antigenicity [6-8]. The major histocompatibility complex (MHC) peptide binding of antigen protein is predicted using neural networks trained on C terminals of known epitopes. In analysis predicted MHC/peptide binding of antigen protein is a log-transformed value related to the IC50 values in nM units. MHC2Pred predicts peptide binders to MHCI and MHCII molecules from protein sequences or sequence alignments using Position Specific Scoring Matrices (PSSMs). Support Vector Machine (SVM) based method for prediction of promiscuous MHC class II binding peptides. SVM has been trained on the binary input of single amino acid sequence [9-14]. In addition, we predict those MHC ligands from whose C-terminal end is likely to be the result of proteosomal cleavage [15].

Results and Interpretations

We found binding of peptides to a number of different alleles

using Position Specific Scoring Matrix. An antigen protein sequence is 426 residues long, having antigenic MHC binding peptides. MHC molecules are cell surface glycoproteins, which take active part in host immune reactions and involvement of MHC class-I and MHC II in response to almost all antigens. PSSM based server predict the peptide binders to MHCI molecules of antigen protein sequence are as 11mer_H2_Db, 10mer_H2_Db, 9mer_H2_Db, 8mer_H2_Db and also peptide binders to MHCII molecules of antigen protein sequence as I_Ab.p, I_Ad.p, analysis found antigenic epitopes region in putative antigen protein (Table 1). We also found the SVM based MHCII-IAb peptide regions; MHCII-IAd peptide regions; MHCII-IAg7 peptide regions and MHCII- RT1.B peptide regions, which represented predicted binders from bacterial antigen protein (Table 2). The predicted binding affinity is normalized by the 1% fractil. We describe an improved method for predicting linear epitopes (Table 2). The region of maximal hydrophilicity is likely to be an antigenic site, having hydrophobic characteristics, because terminal regions of antigen protein is solvent accessible and unstructured, antibodies against those regions are also likely to recognize the native protein (Figure1, 2, 3). It was shown that a antigen protein is hydrophobic in nature and contains segments of low complexity and high-predicted flexibility (Figure 4, 5). Predicted antigenic fragments can bind to MHC molecule is the first bottlenecks in vaccine design.

Conclusion

An antigen protein from *Echinococcus multilocularis* peptide nonamers are from a set of aligned peptides known to bind to a given MHC molecule as the predictor of MHC-peptide binding. MHCII molecules bind peptides in similar yet different modes and alignments

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MHC-I	POS.	N	Sequence	C	MW (Da)	Score	% OPT.
8mer_H2_Db	174	SSL	SDVTPRYI	EDM	932.05	19.64	37.41 %
8mer_H2_Db	300	DRG	GPIPGTTD	SGS	738.79	13.544	25.80 %
8mer_H2_Db	280	LAD	GSSGGVTL	TSL	658.7	11.871	22.61 %
8mer_H2_Db	219	SAK	PNLSQIFV	GLR	899.06	10.214	19.46 %
8mer_H2_Db	295	TIT	SPDRGGPI	PGT	779.86	9.066	17.27 %
8mer_H2_Db	60	IAA	GLINPVQL	GIK	835.01	7.939	15.12 %
8mer_H2_Db	374	RPG	VPIRALYD	YVG	928.11	6.218	11.85 %
8mer_H2_Db	343	TSS	VNGAAAAI	SKE	667.76	6.033	11.49 %
8mer_H2_Db	341	DYT	SSVNGAAA	AIS	657.68	5.954	11.34 %
9mer_H2_Db	216	LDI	SAKPNLSQI	FVG	939.08	15.774	31.32 %
9mer_H2_Db	107	YKH	YKKNVRCKK	EYF	1134.35	13.694	27.19 %
9mer_H2_Db	251	YGV	DMAPNFPVF	QEY	1019.19	13.637	27.08 %
9mer_H2_Db	59	NIA	AGLINPVQL	GIK	906.09	12.02	23.87 %
9mer_H2_Db	94	NFD	SEFENAQKT	WYK	1035.08	9.414	18.69 %
9mer_H2_Db	87	TSI	KEVKNFDSE	FEN	1077.16	7.325	14.54 %
9mer_H2_Db	19	LRK	FAARLEMFL	RTG	1079.34	7.156	14.21 %
9mer_H2_Db	390	ADE	LSFNSGDLF	EKL	981.08	7.076	14.05 %
9mer_H2_Db	342	YTS	SVNGAAAAI	SKE	754.84	5.826	11.57 %
10mer_H2_Db	185	EDM	TQVFNKAQAF	ERE	1135.28	16.784	28.52 %
10mer_H2_Db	87	TSI	KEVKNFDSEF	ENA	1224.34	14.513	24.66 %
10mer_H2_Db	14	AYA	SNLRKFAARL	EMF	1157.39	13.843	23.52 %
10mer_H2_Db	275	KGR	SALADGSSGG	VTL	802.8	13.092	22.24 %
10mer_H2_Db	310	DSG	SNISTSPVHT	TAY	1024.09	11.937	20.28 %
10mer_H2_Db	260	PVF	QEYSPEMSAL	GKK	1136.26	10.912	18.54 %
10mer_H2_Db	190	VFN	KAQAFERERI	IYF	1229.42	10.639	18.08 %
10mer_H2_Db	324	AYG	SNSYDHGSEG	ATP	1033.97	10.439	17.74 %
10mer_H2_Db	348	GAA	AAISKEKQRV	EDT	1111.31	10.178	17.29 %
11mer_H2_Db	113	VNR	CKKEYFHACKT	VRS	1339.59	25.822	32.48 %
11mer_H2_Db	410	QGW	CKGRKDGRVGL	YPQ	1170.39	17.044	21.44 %
11mer_H2_Db	59	NIA	AGLINPVQLGI	KNW	1076.3	16.853	21.20 %
11mer_H2_Db	340	SDY	TSSVNGAAAAI	SKE	943.02	16.171	20.34 %
11mer_H2_Db	347	NGA	AAISKEKQRV	EDT	1182.39	14.568	18.33 %
11mer_H2_Db	87	TSI	KEVKNFDSEFE	NAQ	1353.46	9.581	12.05 %
11mer_H2_Db	146	EQL	RKIEDKLRKGI	MEE	1337.63	8.196	10.31 %
11mer_H2_Db	282	DGS	SGGVTLTSLKT	ITS	1045.18	6.949	8.74 %
11mer_H2_Db	185	EDM	TQVFNKAQAFE	RER	1264.4	6.609	8.31 %

Table 1: PSSM based prediction of MHC ligands, from whose C-terminal ends are proteosomal cleavage sites.

MHC ALLELE	Rank	Sequence	Residue No.	Peptide Score
I-Ab	1	RALYDYVGV	377	1.101
I-Ab	2	GVTLTSLKT	284	1.061
I-Ab	3	KTWYKHYKN	101	0.952
I-Ab	4	KGIMEEEKT	154	0.934
I-Ad	1	LSFNSGDLF	390	0.841
I-Ad	2	TATNILSGL	35	0.722
I-Ad	3	TGVEYGTAT	29	0.669
I-Ad	4	GSSGGVTLT	280	0.640
I-Ag7	1	VNGAAAAIS	343	1.876
I-Ag7	2	YGVDMAPNF	248	1.865
I-Ag7	3	YVGVEADEL	382	1.804
I-Ag7	4	SSVNGAAAA	341	1.733
RT1.B	1	KTVRSLQVQ	122	1.377
RT1.B	2	FKEQALQMQ	202	1.230
RT1.B	3	TTAYGSNSY	319	0.947
RT1.B	4	LRKFAARLE	16	0.937

Table 2: SVM based prediction of promiscuous MHC class II binding peptides from antigen protein.

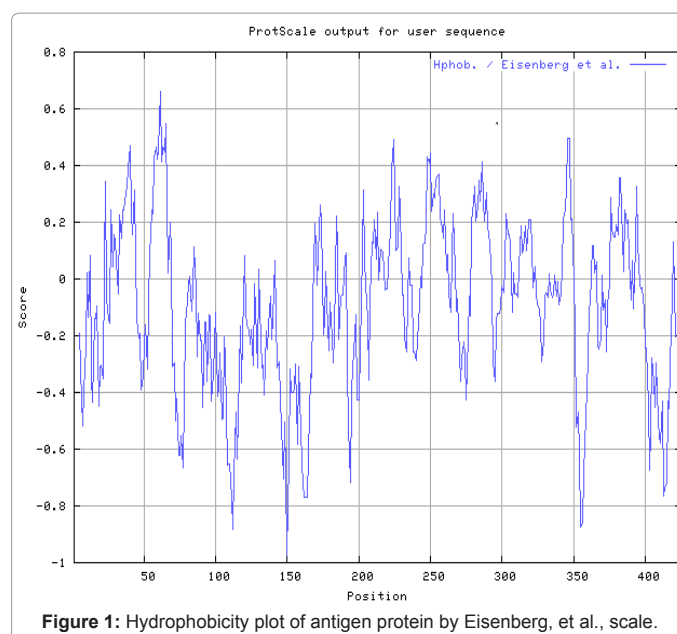
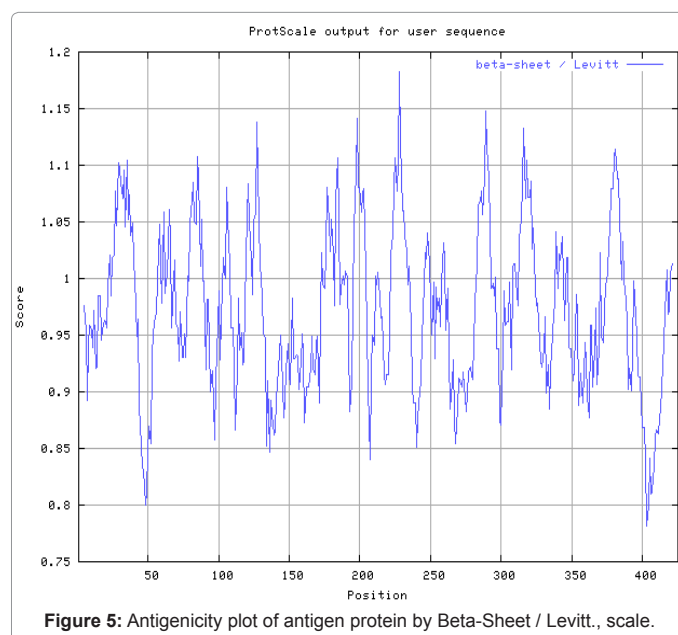
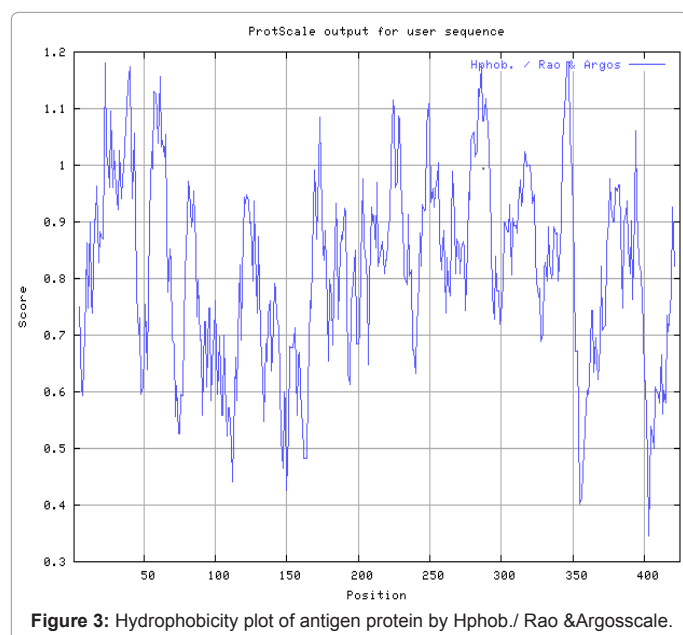
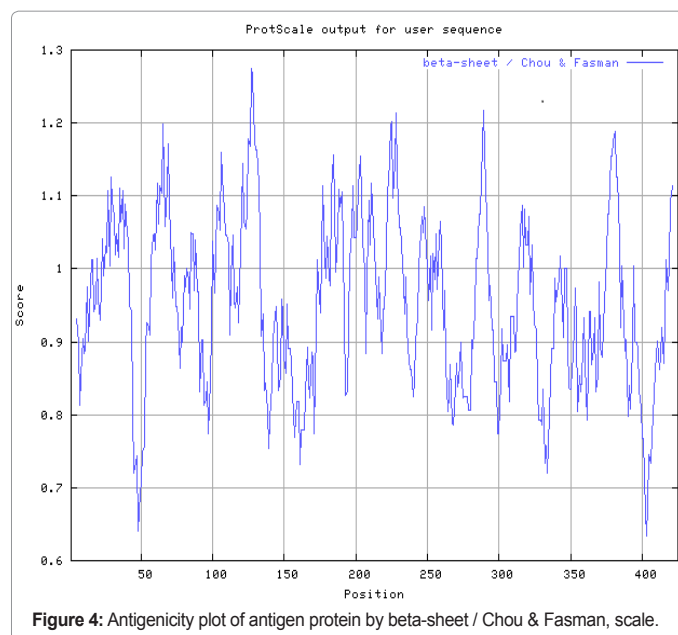
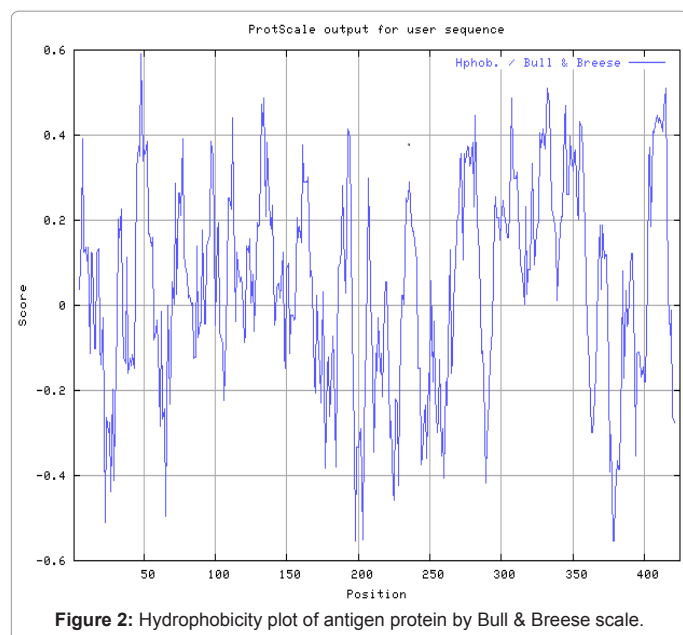


Figure 1: Hydrophobicity plot of antigen protein by Eisenberg, et al., scale.



of MHCII-ligands were obtained to be consistent with the binding mode of the peptides to their MHC class, this means the increase in affinity of MHC binding peptides may result in enhancement of immunogenicity of bacterial antigen protein. These predicted of antigen protein antigenic peptides to MHC class molecules are important in vaccine development from *Echinococcus multilocularis*.

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