1 **Exploring the structural basis to develop efficient multi-epitope vaccines**

2 **displaying interaction with HLA and TAP and TLR3 molecules to prevent**

3 **NIPAH infection, a global threat to human health**

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5 Sukrit Srivastava*[a], Sonia Verma[b], Mohit Kamthania[c], Ajay Kumar Saxena[d],

6 Kailash C Pandey[b], Michael Kolbe*[e,f]

7 a   Infection Biology Group, Indian Foundation for Fundamental Research,
8   RaeBareli 229001, India
9 b   Parasite-Host Biology Group, Protein Biochemistry & Engineering Lab,
10   ICMR-National Institute of Malaria Research, New Delhi, India
11 c   School of Science, SAGE University, M.P. Bhopal, India
12 d   Molecular Medicine Lab., School of Life Science, Jawaharlal Nehru
13   University, 110067 New Delhi, India.
14 e   Department for Structural Infection Biology, Centre for Structural Systems
15   Biology (CSSB) & Helmholtz-Centre for Infection Research, Notkestraße
16   85, 22607 Hamburg, Germany.
17 f   Faculty of Mathematics, Informatics and Natural Sciences, University of
18   Hamburg, Rothenbaumchaussee 19, 20148 Hamburg, Germany.
19
20 *Correspondence should be addressed to S.S (srivastava.sukrit@iffr.in;
21 srivastav.sukrit@gmail.com); and M.K. (Michael.Kolbe@helmholtz-hzi.de)

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23
24 **ABSTRACT**

25 **Background:** Nipah virus (NiV) is an emerging zoonotic virus that caused

26 several serious outbreaks in the South Asian region with high mortality rates

27 ranging from 40 to 90% since 2001. NiV infection causes lethal encephalitis and

28 respiratory disease with the symptom of endothelial cell-cell fusion. No specific

29 vaccine has yet been reported against NiV.

30 **Methodology and Principal Findings:** Recently, the design of some Multi-

31 Epitope Vaccines (MEV) has been proposed but that involves vary limited

32 number of epitopes which limits the potential of vaccine. To address the urgent

need for a specific and effective vaccine against NiV infection, in the present study, we have designed two MEVs composed of 33 Cytotoxic T lymphocyte (CTL) epitopes and 38 Helper T lymphocyte (HTL) epitopes. Both the MEVs carry potential B cell linear epitope overlapping regions, B cell discontinuous epitopes as well as IFN-γ inducing epitopes. Hence the designed MEVs carry potential to elicit cell-mediated as well as humoral immune response. Selected CTL and HTL epitopes were validated for their stable molecular interactions with HLA class I and II alleles as well as in case of CTL epitopes, with human transporter associated with antigen processing (TAP). Human β-defensin 2 and β-defensin 3 were used as adjuvants to enhance the immune response of both the MEVs. Molecular dynamics simulation studies of MEVs-TLR3 ectodomain (Toll-Like Receptor 3) complex indicate the stable molecular interaction. Further, the codon optimized cDNA of both the MEVs has shown high expression potential in the mammalian host cell line (Human). Hence for further studies, the designed MEV constructs could be expressed and tried *in-vivo* as potential vaccine candidates against NiV.

Conclusion: We conclude that the MEVs designed and in silico validated here could be highly potential vaccine candidate to combat NiV, with greater effectiveness, high specificity and large human population coverage worldwide.

Key words: Nipah virus (NiV), zoonotic disease, Human Transporter associated with antigen processing (TAP), Toll-Like Receptor 3 (TLR-3), Epitope,

Immunoinformatics, Molecular Docking, Molecular Dynamics (MD) simulation,

Multi-epitope Vaccine (MEV)

AUTHOR SUMMARY

Nipah Virus (NiV) has caused several outbreaks in past two decades calming

large number of human lives. Our present work aims to design and *in silico*

validate Multi-Epitope Vaccine against NiV. The current approach to design

vaccine involves whole virus or full length proteins as vaccine candidates against

NiV. These approaches carry chances of raising the unwanted non-neutralizing

antibodies which have been found to cause clinical complexities. Recently few

Multi-Epitope vaccines have also been proposed, but they have involved limited

number of epitopes for vaccine design in result limiting the effectiveness and

human population coverage. Here in our MEVs we have involved all the proteins

of NiV to design the vaccine. Moreover since we have used in silico validated

epitopes we may conclude that the here proposed MEVs would be highly

specific, effective and potential vaccine candidate to combat NiV with large

human population coverage worldwide.

INTRODUCTION

Nipah virus (NiV) is an emerging zoonotic virus of the genus Henipavirus

of the Paramyxoviridae family [1]. NiV infection causes fatal encephalitis and

respiratory disease with a particular symptom of endothelial cell-cell fusion [2].

The first NiV infection to human was first reported in Malaysia in 1998. Later NiV

outbreak was reported from Meherpur, Bangladesh in 2001. In the Malaysia NiV

infection, the transmission was primarily due human contact with pigs, whereas

in later outbreaks of Bangladesh and India the transmission was associated with

contaminated date palm sap and human-to-human contact [3]. Bats are identified

as the main reservoir for the NiV and they are responsible for the transmission of

the infection to both humans and animals [4]. After 2001 NiV outbreak has been

reported from different district of Bangladesh almost every year (2003-05, 2007-

12). Till March 31, 2012, a total of 209 confirmed cases of NiV infections were

reported out of which 161 people died resulting in the mortality rate as high as

77%. After several outbreaks in Bangladesh, in total three NiV outbreaks have

also been reported from India. Two of them occurred in the state of West Bengal

in 2001 and 2007 [5]. The most recent NiV outbreak was reported from the

Kerala state of India during the period of May to June-2018. The Kerala outbreak

claimed 17 lives leaving only two survivors out of 19 confirmed cases [6]. Till

present, there has been no specific vaccine reported against NiV infection, and

the pathogenesis mechanism of NiV to human cells is largely unknown. Hence,

an immune-informatics approach investigating the potential of different NiV

proteins for vaccine design would be an important and essential step forward for

vaccine development.

NiV infection of human cells involves several protein-protein interactions

and protein cluster formation on the host cell surface. Essential proteins involved

in NiV pathogenesis include C protein, Fusion glycoproteins (F), Glycoproteins

(G), Matrix proteins (M), Nucleocapsid protein, Phosphoprotein, Polymerase, V

protein and the W protein [7-21]. The C protein regulates the early host pro-

inflammatory response as well as the pathogen virulence thus providing a

conducive environment for a successful NiV infection [7]. The attachment

glycoprotein (G), the fusion protein (F) and the matrix protein (M) together form a

cluster on the human cell membrane facilitating virus particle assembly and

pathogenesis [8-12]. The G and F proteins of NiV have been shown to be

immunogenic by inducing protective immune responses in hamsters [22, 23]. The

NiV matrix protein is observed to play a central role in virus particle formation and

is essentially required for viral budding from the infected human cells [13-15].

The NiV Polymerase is responsible for the initiation of RNA synthesis, primer

extension, and transition to elongation mode and hence the enzyme facilitates

viral pathogenesis and survival in host cells [16]. The phosphoprotein and the

glycoprotein of NiV are crucially involved in the regulation of viral replication [17,

18] while the V protein of NiV is responsible for the host interferon (IFN) signaling

evasion during pathogenesis [19, 20]. Interestingly, the identical N-terminal

region of the pathogen's V and W protein is sufficient to exert the IFN-antagonist

activity [21]. Hence, all the nine above mentioned NiV proteins are crucial in

different ways for viral pathogenesis and might be important drug and vaccine

candidates.

The Nipah virus is a zoonotic RNA virus and it infects human respiratory

epithelium cells as well as differentiated neurons (in the brain and spinal cord).

Thus, as understood by previous animal model studies, recovery from viral

infection and the clearance of viral RNA requires the presence of virus-specific

antibodies and interferon gamma (IFN-γ) secretion from T cells [24-26]. Along

with the B cell, the T cell also play a critical role in immune response against NiV

infection. In recent studies a number of B cell and T cell epitopes from the NIPAH

proteome have been reported [27-38]. Further different approaches were

proposed for the design of multi epitope vaccines [39, 44]. However, the

proposed vaccines utilized a very limited number (6 to 8) of T and B cell

epitopes. The use of limited number of epitopes could be challenging for the

successful presentation of the exogenous vaccine candidates in view of the

proteolytic cleavage by Antigen Presenting Cells (APC).

In the present study we have screened out the most potential Cytotoxic T

lymphocyte (CTL) epitopes, Helper T lymphocyte (HTL) and B cell epitopes from

the NiV proteome. We have shortlisted and prioritized the most potential and

highest scoring 33 CTL, 38 HTL and 16 B cell epitopes. We further evaluated

several critical properties (like IC(50), Immunogenicity, Conservancy, Non-toxic

etc) to identify the epitopes with the highest immunogeneic potential against NiV.

The shortlisted epitopes were utilized for the design of CTL and HTL multi-

epitope vaccines against NiV. The designed vaccines were further studied for

their stable interaction with innate immunity Toll-Like Receptor 3 (TLR3). The

analysis of cDNA of the designed multi-epitope vaccine has predicted to be

highly favorable for expression in mammalian cell line. Overall in the present

study we have designed and proposed potential multi-epitope vaccines against

Nipah virus infection.

RESULTS & DISCUSSION

Screening of potential epitopes

T cell Epitope Prediction

Screening of Cytotoxic T lymphocyte (CTL) Epitope. Cytotoxic T lymphocyte

(CTL) epitopes screened and shortlisted according to the highest "Total Score",

low IC(50) (nM) value for epitope-HLA class I allele complexes, and epitopes with

the larger number of the HLA class I allele binders. The immunogenicity of the

shortlisted CTL epitopes was also determined; the higher immunogenicity score

indicates the greater immunogenic potential of the given epitope (Supplementary

table S3, S7). In total 33 CD8+ T cell epitopes were finally selected. From this list

10 CTL epitopes (Fusion Protein: FALSNGVLF;   Glycoprotein:

TVYHCSAVY; Nucleocapsid: YPALALNEF;   Phosphoprotein:

VSDAKMLSY; Polymerase: YPECNNILF,   FPVMGNRIY,

AEFFSFFRTF, IPFLFLSAY, ETDDYNGIY, SQNLLVTSY) show a match with

previous studies [35-38], indicating consensus of epitope screening by different

approaches and methods (Supplementary table S3).

Screening of Helper T lymphocyte (HTL) epitopes. The screening of helper T

lymphocyte (HTL) epitopes from nine different proteins of NiV was performed on

the basis of "Percentile rank". The smaller the value of percentile rank the higher

would be the affinity of the peptide with its respective HLA allele binders. In our

initial screening, we got several potential CD4+ T cell epitopes with high scoring.

38 epitopes out of initial screening were shortlisted on the basis that they had

highest percentile rank and highest number of HLA class II allele binders

(Supplementary table S4, S8).

Population Coverage by CTL and HTL epitopes. The population coverage by the

shortlisted epitopes was also studied, in particular involving countries of South

Asia, East Asia, Northeast Asia and the Middle East. From this study, we

concluded that the combined use of all the shortlisted CTL and HTL epitopes

would have a cumulative percent of world (countries as listed in Supplementary

table S5) population coverage as high as 97.88%.

B Cell epitope prediction

Sequence-based B Cell epitope prediction. In our initial study, we screened a

total of 116 B Cell epitope from nine different NiV proteins, with the epitope

length of at least four amino acids utilizing the Bepipred Linear Epitope Prediction

method. B cell epitopes predicted by another five different methods based on

different physiochemical properties were found to have significant consensus

with the epitope amino acid sequences predicted by Bepipred Linear Epitope

Prediction. Here, 16 out of the 116 epitopes were shortlisted having a length of 4

to 19 amino acids (Supplementary table S6, Figure 1). One of these 16 B cell

epitopes (Matrix Protein: SIPREFMIY) matches with a previous study [39],

indicating epitope screening consensus using different approaches and methods

(Supplementary table S6).

Figure 1. Overlapping regions amongst the linear B cell epitopes predicted by the BepiPred method and other seven different B cell epitopes prediction methods. B cell epitopes predicted by the BepiPred method and other different protein sequence based (Chou.., Emini.., Karplus.., Kolaskar.., and Parker..) and protein structure based (DiscoTope and ElliPro) prediction methods were found to have significant consensus. Consensus overlapping regions of BepiPred epitopes are underlined by the different color, corresponding to respective prediction method.

Structure-based B cell epitope prediction. Structure-based discontinuous and linear epitopes predicted by the DiscoTope 2.0 and the Ellipro methods have shown a significant consensus of overlapping amino acid sequence with the linear epitopes predicted by Bepipred linear epitopes method (Supplementary table S6, Figure 1). This result further confirms that the shortlisted B Cell Bepipred Linear Epitopes have a high chance of causing a strpmg immuno response.

Characterisation of potential epitopes

Epitope conservation and toxicity analysis. Sequence conservation analysis of the shortlisted 33 CTL, 38 HTL and 16 B cell epitopes show that the amino acid sequence conservancy of CTL, HTL and B cell epitopes amongst all retrieved NiV protein sequences is mostly 100%, as shown in Supplementary table S3, S4 and S6. Further, the toxicity analysis of all the shortlisted CTL, HTL and B Cell

epitopes was also performed. The ToxinPred study indicated the non-toxic nature

of all the shortlisted epitopes (Supplementary table S3, S4, S6).

Overlapping residue analysis. Analysis of the amino acid sequence overlap

amongst the shortlisted CTL, HTL and B cell epitopes from nine NiV proteins was

performed by using the Multiple Sequence Alignment (MSA) analysis tool Clustal

Omega. Our analysis showed that several epitopes of CTL, HTL and B cell have

overlapping amino acid sequence. Besides a few shorter identical sequences the

majority of the CTL, HTL and B cell epitopes have several amino acids

overlapping as shown in Figure 2.

Figure 2. Overlapping CTL, HTL and B cell epitopes. Multiple sequence

alignment performed by Clustal Omega at EBI to identify the consensus

overlapping regions of CTL (red), HTL (blue) and B cell epitopes (green)

amongst shortlisted epitopes. Epitopes with overlapping region amongst all the

three types of epitopes (CTL, HTL and B Cell epitopes), epitopes with full

sequence overlap and epitopes with the highest number of HLA allele binders

were selected for further studies (encircled).

Epitope selected for molecular interaction study with HLA allele and TAP

transporter. Amongst all the shortlisted epitope peptides seven CTL and

seventeen HTL epitope peptides have partial or full overlapping sequences or

have the highest number of HLA allele binders were shortlisted for further studies

(Figure 2, Supplementary table S3 & S4).

Molecular interaction analysis of selected epitopes with HLA allele and TAP

transporter.

Molecular interaction analysis of selected CTL and HTL epitopes with HLA

alleles. The molecular docking study of the shortlisted CTL and HTL epitopes

with their respective HLA class I and II allele binders was performed using

AutoDock Vina. Docking studies revealed for all epitopes significant molecular

interactions with their HLA allele binders having low binding energies and

multiple hydrogen bonds formed (Figure 3A & 4B). The stability of the obtained

docking complexes was further tested by molecular dynamics (MD) simulation

studies. MD simulations were performed over a time interval of 0.5-1 ns at the

invariable temperature of ~ 300 K and at invariable pressure of ~ 1 bar. All the

complexes showed reasonably invariant root mean square deviation (RMSD)

value (between ~ 0.2 to 0.4 nm) indicating the stable nature of the tested epitope-

HLA allele complexes (Figure 4A & 5B). Moreover, the reasonably invariant Rg

(radius of gyration) of the complexes, throughout the MD simulation

(Supplementary figure S2), and the root mean square fluctuation (RMSF) for all

the atoms of the complexes (Supplementary figure S3) again indicate the stable

nature of the epitopes and HLA allele complexes. Furthermore, the B-factor

analysis of all the epitope-HLA allele complexes indicated most of the complex

regions to be stable (blue) with a very small region being acceptably fluctuating

(yellow and orange) (VIBGYOR color presentation) Supplementary figure S4.

Figure 3. (A) Molecular Docking analysis of CTL epitopes and HLA alleles.

Molecular docking of selected CTL epitopes (cyan sticks) with their respective

HLA class I allele binders (gray sticks). The study shows the docked complexes

to have significantly negative binding energy along with hydrogen bonds (green

dots) formation in the complex interface. **(B) Molecular Docking analysis of**

HTL epitopes and HLA alleles. Molecular docking of selected HTL epitopes

(cyan sticks) with their respective HLA class II allele binders (gray sticks). The

study shows the docked complexes to have significantly negative binding energy

along with hydrogen bonds (green dots) formation in the complex interface. (*)

Indicates binding energy, shown in kcal/mol.

Figure 4. (A) Molecular Dynamics simulation analysis of CTL epitopes and

HLA allele complexes. Molecular Dynamics simulation study reveals a stable

nature of the CTL-HLA allele complexes throughout 0.5-1 ns time window with

reasonably invariable RMSD. **(B) Molecular Dynamics simulation analysis of**

CTL epitopes and HLA allele complexes. Molecular Dynamics simulation study

reveals a stable nature of the HTL-HLA allele complexes throughout 0.5-1 ns

time window with reasonably invariable RMSD.

Molecular interaction analysis of selected CTL epitopes with TAP transporter.

The molecular docking interaction analysis of the chosen CTL epitopes with the

TAP transporter cavity showed a strong molecular interaction with low binding

energy and several hydrogen bonds formed at different sites of the TAP

transporter cavity. Two sites of interaction were of particular interest, one located

near the cytoplasmic end and the other in the vicinity of the ER lumen (Figure 5).

Our study confirms the transportation feasibility of the chosen CTL epitopes from

the cytoplasm into the ER lumen which is essential for the representation of

peptides by the HLA allele molecules on the surface of antigen presenting cells.

Figure 5. Molecular docking analysis of CTL epitopes within the TAP

transporter cavity. Molecular interaction of the seven selected CTL epitopes

(cyan sticks) with the TAP cavity (gray ribbon/sticks) is shown in detail. Panel **(A)**

shows the binding of epitope at two different sites within TAP cavity, panel **(B)**

and **(C)** show detailed molecular interaction between epitopes and TAP cavity;

(a, b) show chain A and B of TAP transporter. H-bonds are highlighted as yellow

dots. (*) Indicates binding energy, shown in kcal/mol.

Design, characterisation and molecular interaction analysis of Multi-

Epitope Vaccines with immunological receptor.

Design of Multi-Epitope Vaccines. All the screened CTL and HTL epitopes were

utilized to design CTL and HTL Multi-Epitope vaccines. Short peptides EAAAK

and GGGGS were used as rigid and flexible linkers respectively (Figure 6). The

GGGGS linker provides proper conformational flexibility to the vaccine tertiary

structure and hence facilitates stable conformation to the vaccine. The EAAAK

linker facilitates in domain formation and hence facilitates the vaccine to obtain

its final stable structure [45-54]. The human beta-defensin 2 (hBD-2) (PDB ID:

1FD3, Sequence: GIGDPVTCLKSGAICHPVFCPRRYKQIGTCGLPGTKCCKKP)

and the human beta-defensin 3 (hBD-3) (PDB ID: 1KJ6, Sequence:

GIINTLQKYYCRVRGGRCAVLSCLPKEEQIGKCSTRGRKCCRRKK) were used

as adjuvants in the design of both the MEVs at N and C terminals respectively

[45-54,]. Human Beta-defensins (hBD) have an important role in the chemotactic

activity memory T cells, immature dendritic cells and monocytes and they are

involved in the degranulation of mast cells. Due to the important role of hBDs in

immune response enhancement, hBDs have been chosen and utilized as

adjuvants for the MEV designs.

Figure 6. Design of Multi-Epitope Vaccine (MEVs). (A) CTL and **(B)** HTL

epitopes were linked by the short peptide linker 'GGGGS'. Human β Defensin 2

and β Defensin 3 were used as an adjuvant at the N and C terminals

respectively. The short peptide EAAAK was used to link the β Defensin 2 and β

Defensin 3. Epitopes from different proteins were highlighted with different colors

and the C terminal 6xHis is designated as His tag. *Indicates the epitopes

common to Phosphoprotein, V Protein and W protein, dues to protein sequence

similarity.

Characterisation and molecular interaction analysis of designed Multi-Epitope Vaccines with immunological receptor

Characterisation of designed Multi-Epitope Vaccines

Interferon-gamma inducing epitope prediction. Interferon-gamma (IFN-γ) inducing epitopes are involved in both the adaptive and the innate immune response. The IFN-γ inducing 15-mer peptide epitopes were screened from the CTL and HTL MEVs by utilizing the IFNepitope server. A total of 33 CTL MEV and 43 HTL MEV INF-γ inducing POSITIVE epitopes with a score of 1 or more than 1 were shortlisted (Supplementary table S9, Figure 7D & 7I).

Figure 7. Tertiary structure modelling of CTL and HTL Multi-Epitope Vaccines. (A) & (F): Tertiary structural models of CTL and HTL MEVs showing epitopes in different colors corresponds to as in Figure 6. **(B) & (G)**: Show the different domains of CTL and HTL MEVs. **(C) & (H)**: The overlapping linear B cell epitope region present in CTL and HTL MEVs, shown by spheres. **(D) & (I):** From the CTL and HTL MEVs, the INF-γ inducing epitopes are shown in cyan, discontinuous B Cell epitopes are shown in magenta and the region common amongst INF-γ and discontinuous B Cell epitopes are shown in wheat color. **(E) & (J)**: RAMPAGE analysis of the refined CTL and HTL MEV models.

MEVs allergenicity and antigenicity prediction. Both the CTL and HTL MEVs were analyzed to be NON-ALLERGEN by the AlgPred analysis scoring -0.612 and -0.935 respectively while default threshold value being -0.4. CTL and HTL

MEVs were also analyzed by VaxiJen to be probable ANTIGENS with the

prediction score of 0.4447 and 0.4836 respectively, while the default threshold

value for viral proteins being 0.4. Hence, with the mentioned analysis tools both

the CTL and HTL MEVs are predicted to be non-allergic as well as antigenic in

nature.

Physicochemical property analysis of designed MEVs. ProtParam calculations

were performed for both the CTL and HTL MEVs to analyse their physiochemical

properties. The CTL MEV is composed of 576 amino acids, has a molecular

weight of 58.57 kDa and a theoretical pI of 8.19. The expected half-life of the

CTL MEV in *E.coli*, yeast and mammalian reticulocytes were predicted with 10 h,

20 min, and 30 h respectively; the aliphatic index of CTL MEV was found to be

58.42, and grand average of hydropathicity (GRAVY) of CTL MEV was found to

be -0.010, both indicating globular and hydrophilic nature of the CTL MEV. The

instability index score of the CTL MEV was 48.03 indicating its stable folding

under native conditions. Further, the ProtParam analysis of the HTL MEV

calculated for the 857 amino acids, a molecular weight of 87.62 kDa and a

theoretical pI of 5.99. The expected half-life of HTL MEV in *E.coli*, yeast and

mammalian reticulocytes was predicted to be 10 h, 20 min and 30 h,

respectively. The aliphatic index of HTL MEV was calculated as 82.99, and the

grand average of hydropathicity (GRAVY) of the HTL MEV was found to be

0.188, both indicating that HTL MEV has a globular and hydrophilic nature. The

instability index of the HTL MEV was 45.66 indicating its stable nature.

Overall, the physiochemical properties of both proteins indicate that both

the proteins might be natively expressed as stable proteins and purified with

standardized purifications methods and protocols.

Tertiary structure modelling and refinement of MEVs. Tertiary structural models

were generated for both the CTL and HTL MEVs by utilizing the RaptorX

modelling tool (Figure 7). The model obtained for CTL MEV has 6% helix, 27% β-

sheet, 66% coil content and structural elements are 23% exposed, 39% medium

and 36% buried. The structural model of CTL MEV has three domains ranging

from amino acid 1 to 46 (1st domain, template-1fd3:A), 47 to 520 (2nd domain,

templates-1yrzA, 1y7bA, 5jozA, 3zxjA, 5z5dA) and 521 to 576 (3rd domain ,

template-1kj6:A) (Figure 7B). Similarly, the 3D model calculated with RaptorX for

the HTL MEV has 21% helix, 22% β-sheet, 56% coil content with 24% of the

amino acids exposed, 39% medium and 36% buried. The structural model of

HTL MEV has three domains ranging from amino acid 1 to 46 (1st domain,

template-1fd3:A), 47 to 801 (2nd domain, templates-5m5zA, 3eqnA), 802 to 857

(3rd domain, template-1kj6:A) (Figure 7G). The P-Value for the best template

based CTL and HTL MEV homology models were 2.79e-04 and 5.99e-03

respectively. Good quality, mostly alpha proteins have a P-value of less than 10^{-3}

and that of mostly beta proteins has a P-value of less than 10^{-4}. Hence, both the

homology models of CTL and HTL MEVs are predicted to be of good quality.

Since for the CTL and HTL MEV design, the CTL and HTL epitopes used also

show overlapping common regions with the linear B cell epitopes (Figure 1), both

the generated CTL and HTL MEV models also carry the overlapping regions of

linear B Cell epitopes (Figure 7C, 7H).

399   The generated CTL and HTL 3D models were further refined using

ModRefiner to fix structural gaps followed by GalaxyRefine refinement.

Refinement with ModRefiner showed a TM-score of 0.9703 and 0.8934 for the

CTL and HTL models, respectively. Since both values are close to 1 this

indicates that the initial and the refined models were structurally similar. For the

CTL MEV model refinement, the sores of model 1 for different parameters were,

Rama favored: 90.8%, GDT-HA: 0.9596, RMSD: 0.385, MolProbity: 2.673, Clash

score: 29.9, and Poor rotamers: 1.8. Likewise for HTL MEV model refinement,

the sores of model 1 for different parameters were, Rama favoured: 88.5%, GDT-

HA: 0.9463, RMSD: 0.419, MolProbity: 2.811, Clash score: 38.4, and Poor

rotamers: 1.6. Here, MolProbity shows the log-weighted combination of the clash

score, percentage Ramachandran not favoured and the percentage bad side-

chain rotamers. After refinement, all the mentioned parameters were significantly

improved in comparison to the initial CTL and HTL MEV models (Supplementary

table S10).

Validation of CTL and HTL MEVs refined models. Both the CTL and HTL model

were analysed with the RAMPAGE analysis tool after refinement. The refined

CTL MEV model has 91.5% residues in favored regions and the refined HTL

MEV model has 89.5% of residues in favored region. This indicates both the CTL

and HTL MEVs have gained an acceptable conformation after refinement (Figure

7E, 7J).

Discontinuous B-cell epitope prediction from MEVs. Discontinuous B-cell

epitopes were further predicted from the final refined 3D models of CTL and HTL

MEVs utilizing the ElliPro tool on IEDB server. The screening revealed that the

CTL MEV carries 3 and the HTL MEV has 2 potential discontinuous epitopes.

The PI (Protrusion Index) score of the CTL MEV discontinuous B cell epitopes

ranges from 0.682 to 0.747 and that of HTL MEV it ranges from 0.687 to 0.745

(Supplementary table S11, Figure 7D & 7I). The higher PI score indicates a

greater potential of the discontinuous B cell epitope.

Molecular interaction analysis of MEVs with TLR-3 .

The refined models of CTL and HTL MEVs were further studied for their

molecular interaction with the ectodomain (ECD) of human TLR-3. Therefore,

molecular docking of CTL and HTL MEVs model with the TLR-3 crystal structure

model (PDB ID: 2A0Z) was performed utilizing the PatchDock tool. Generated

docking conformation with highest geometric shape complementarity score of

22382 and 18264 for CTL and HTL MEVs, respectively were selected for further

studies. The highest docking score predicted with the PatchDock tool indicates

the best geometric shape complementarity fitting conformation of MEVs and the

TLR-3 receptor. Both, the CTL and HTL MEVs were fitting into the ectodomain

region of TLR-3 after docking (Figue 8A & 8E). The CTL and HTL MEVs have

shown to form multiple potential hydrogen bonds within the ectodomain cavity

region of TLR-3. Further, the molecular dynamics simulation study was also

performed for the docked complexes of both the MEVs and TLR-3. In MD

simulations both the complexes have shown reasonably stable effective RMSD

value between ~ 0.2 to 4 Å for a given time window of 100 ns at invariable

pressure (~ 1 bar) and temperature (~ 300 K) (Figure 8B & 8F). The effective

RMSD is well within tange of Hydrogen bond formation, moreover the range is

towards 0.2 Å for most of duration. The reasonably invariant radius of gyration

(Rg) (~ 0.7 to 2 Å) of both MEVs-TLR-3 complexes (Figure 8C & 8G), and RMS

fluctuation (RMSF) (~ 2 to 3 Å) for all the atoms in both the complexes (Figure 8D

& 8H) indicate that the MEVs-TLR3 complexes are stable. The B-factor analysis

of MEVs-TLR3 complexes was also performed. The B-factor indicates the

displacement of the atomic positions from an average (mean) value i.e. the more

flexible an atom is the larger the displacement from the mean position will be

(mean-squares displacement) (Figure 8A and 8E). The areas with high B-factors

are colored red (hot), while low B-factors are colored blue (cold) (VIBGYOR

presentation). The B-factor of most of the regions of MEVs-TLR3 complexes

indicates the stable nature of the complexes while a very small region is found to

be fluctuating. The results suggest a stable complex formation tendency for both

the CTL and HTL MEVs with the ectodomain of the human TLR-3 receptor.

Figure 8. Molecular Docking and dynamics simulation study of CTL and

HTL MEVs with TLR-3. (A) CTL and **(E)** HTL MEVs (VIBGYOR) docked

complex with TLR-3 (gray). Both the complexes are forming several hydrogen

bonds in the MEV and TLR-3 interface, as shown by green dots. B-Factor of the

docked MEVs is shown by a rainbow (VIBGYOR) presentation. The regions in

blue being indicated stable and the region in red indicate unstable. In the above

complexes, most of the region of docked MEVs is in blue and with the very small

region is green, yellow or orange, hence the complexes are predicted to be very

stable. **(B)** and **(F),** RMSD as generated by the molecular Dynamics simulation

study of CTL, HTL MEVs and TLR-3 complexes. **(C)** & **(G)** Rg (radius of gyration)

across the time window of 100 nanosecond. **(D)** & **(H)**, RMS fluctuation for all the

atoms of the CTL, HTL MEVs and TLR-3 complexes.

***In-silico* analysis of MEVs for cloning and expression**

Analysis of cDNA of both the MEVs for cloning and expression in mammalian cell

line. Complementary DNA codon optimized for CTL and HTL expression in

mammalian host cell lines (Human) was generated with the Java Codon

Adaptation Tool. The generated optimized cDNA's for both the MEVs were also

analysed by utilizing the GenScript Rare Codon Analysis Tool revealing aGC

content of optimized CTL-MEV cDNA of 69.79% and a CAI (Codon Adaptation

Index) score of 1.00 with 0% tandem rare codons. Likewise, the GC content of

the optimized HTL-MEV cDNA was 70.69%, CAI score was 1.00 with 0% tandem

rare codons. Since for higher possibility for cDNA expression in human

expression system, the GC content of a cDNA should be within the range of 30%

to 70%, the CAI score should be between 0.8-1.0, and the tandem rare codon

frequency that indicates the presence of low-frequency codons, should be <30%,

the cDNA constructs of CTL and HTL MPVs are expected to have high potential

for expression in human expression system. The tandem rare codons may hinder

the proper expression of the cDNA or even interrupt the translational machinery

of the chosen expression system. According to the GenScript Rare Codon

Analysis the cDNA of both the MEVs satisfy all the mentioned parameters and

are predicted to have high expression in the mammalian host cell line (Human).

CONCLUSION

In the present study, we have designed and validated two multi-epitope

vaccines derived from CTL and HTL epitopes. The selected peptides show

significant sequence overlap with screened linear B cell epitopes. Both, the CTL

and HTL MEVs tertiary models carry potential discontinuous B cell epitopes and

INF-γ epitopes. Consequently, the designed MEVs might have the potential to

elicit profound humoral and cellular immune responses. Human β-Defensin 2 and

3 fused to the N and C terminal ends of both the MEVs serving as adjuvants to

enhance the immune response. The identified epitopes for the CTL and HTL

MEVs were also validated by molecular docking and MD simulation studies to

test the interaction with their respective HLA allele binders. Molecular interaction

of the selected CTL epitopes within the TAP transporter cavity was also

evaluated indicating a favorable transport of epitopes from cytoplasm to lumen of

Endoplasmic Reticulum for further presentation on cell surface by Golgi bodies.

Our analysis of the shortlisted CTL and HTL epitopes combined revealed

coverage of 97.88% world human population. The molecular interaction analysis of both the CTL and HTL MEVs with the immunoreceptor TLR3 showed structural fit of the MEVs into the ectodomain of TLR3 receptor cavity and MD simulations indicate very stable complexes formed. Since both the CTL and HTL MEVs carry CTL, HTL as well as discontinuous B cell epitopes, the combined administration of both the MEVs, is predicted to elicit both the humoral as well as cell-mediated immune response. The cDNA for both the MEVs were designed considering codon-bias for the expression in mammalian host cell lines (Human). The cDNAs were optimized with respect to their GC content and zero tandem rare codons to gain high expression in mammalian host cell lines (Human). In future experimental studies the designed CTL and HTL MEVs should be cloned, expressed and validated *in-vivo* and in animal trails as potential vaccine candidates against NiV infection.

METHODOLOGY

In the present study, we have designed two multi-epitope vaccines (MEVs) composed of thoroughly screened most potential Cytotoxic T lymphocyte (CTL) epitopes and Helper T lymphocyte (HTL) epitopes derived from the nine NiV proteins (glycoprotein (gi-253559848); C protein (gi-1859635642); fusion protein (gi-13559813); matrix protein (gi-13559811); nucleoprotein (gi-1679387250); phosphoprotein (gi-1802790259); V protein (gi-1802790260); RNA polymerase (gi-15487370); W protein (gi-374256971). CTL and HTL epitopes would be the most potential vaccine candidates since they are responsible for

cell-mediated immune response by their presentation on the surface of antigen

presenting cells (APCs) by their respective Class I and II human leukocyte

antigen (HLA) allele binders. Both the CTL and HTL epitopes chosen for MEV

design also carry overlapping regions of linear B cell epitopes. Moreover, both

the MEVs also carry potential discontinuous B cell epitopes as well as IFN-γ

inducing epitopes in their tertiary structure model. Hence, both the designed

MEVs carry potential to elicit cell-mediated as well as humoral immune response.

Furthermore, both the MEVs were designed with the vaccine adjuvant human β

Defensin 2 and human β Defensin 3 at their N and C termini to enhance the

immune response [45-46]. β-defensins have considerable immunological

adjuvant activity in general and, in particular the β-defensins 2 & 3 have been

shown in previous studies to generate potent humoral immune responses when

fused with B-cell lymphoma epitopes in mouse model [47-50]. Since, the pro-

inflammatory mediators enhance the expression of β-defensins 2 & 3 in airway

epithelial cells we chose β-defensins in our study. The selected CTL and HTL

epitopes were validated for their stable molecular interactions with their

respective HLA alleles binders; for CTL epitopes, their molecular interaction with

human Transporter associated with antigen processing (TAP) was also analyzed

[55, 56]. This analysis validated the CTL epitopes that get transported through

the TAP cavity for their presentation on cell surface or not. The human TAP

selectively transports cytosolic peptides into the lumen of the endoplasmic

reticulum in an ATP-dependent manner [57]. Tertiary structure models of both

the MEVs were generated, refined, and further docked with the human Toll-Like

Receptor 3 (TLR3), which is an essential immuno-receptor in this pathway [58,

59]. The nuclear localization of Nipah virus W protein inhibits the signaling

pathway of TLR3 upon pathogenesis to suppress the TLR3 induced activation of

the IFN response to eventually prevent relay of the warning signals to uninfected

cells. TLR3 is preferentially expressed by human astrocytes of the central

nervous system (CNS) upon infection. The NiV infection involves invasion of the

neurons of CNS and hence causes infection in CNS. These studies indicate the

importance of the TLR3 responses in immune response and hence TLR3 has

been chosen to be studied for its stable binding with the designed multi-epitope

vaccines [60-63]. The complexes of CTL and HTL MEVs formed with the human

TLR3 were further analysed for their stable molecular interaction by a molecular

dynamics simulation study. The cDNA of the designed MEVs were generated

and analysed for their high expression tendency in the mammalian host cell line

(Human). Overall, from the present *in-silico* study, we may put forward the design

of two MEVs, which qualify all the significant criterions for being a potential

vaccine candidate against NiV infections. The corresponding workflow is shown

in Supplementary figure S1.

NiV proteins selected for potential epitope screening. In the present study, nine

NiV proteins were used for epitope screening. They include C protein, Fusion

glycoprotein (F), Glycoprotein (G), Matrix proteins (M), Nucleocapsid protein,

Phosphoprotein, Polymerase, V protein and the W protein. The full-length protein

sequences of NiV proteins were retrieved from the NCBI database (National

Center for Biotechnology Information (https://www.ncbi.nlm.nih.gov/protein). A

total of 161 protein sequences available at NCBI, belonging to different strains

and origins of NiV, were retrieved. For structural based epitope screenings

available tertiary structures of NiV proteins were retrieved from Protein Data

Bank (PDB) (http://www.rcsb.org/pdb/home/home.do). For the NiV proteins with

no structure available, homology models were generated by Swiss-model,

(http://swissmodel.expasy.org/) [64] (Supplementary table S1).

Screening of Potential Epitopes

T cell Epitope Prediction

Screening of Cytotoxic T lymphocyte (CTL) Epitope. The screening of Cytotoxic

T lymphocyte epitopes was performed by the IEDB (Immune Epitope Database)

tool "Proteasomal cleavage/TAP transport/MHC class I combined predictor"

(http://tools.iedb.org/processing/) [65-67]. Proteasome cleavage score depends

on the total amount of cleavage site in the protein. TAP score estimates an

effective log -(IC50) values (half maximal inhibitory concentration (IC50)) for

binding to TAP of a peptide or its N-terminal prolonged precursors. The MHC

binding prediction score is -log(IC50) values for binding to MHC of a peptide [68].

The tool gives a "Total Score" which is a combined score of the proteasome,

MHC, TAP (N-terminal interaction), and processing analysis scores. The total

score is generated by using the combination of six different methods viz.

Consensus, NN-align, SMM-align, Combinatorial library, Sturniolo and

NetMHCIIpan [65-68]. The IC(50) (nM) value for each epitope and MHC allele

binding pairs were also obtained by this IEDB tool. Epitopes having high,

intermediate, and least affinity of binding to their HLA allele binders have IC50

values < 50 nM, < 500 nM and < 5000 nM respectively. Immunogenicity of all the

screened CTL epitopes was also obtained by using "MHC I Immunogenicity" tool

of IEDB (http://tools.iedb.org/immunogenicity/) with all the parameters set to

default analyzing 1st, 2nd, and C-terminus amino acids of the given screened

epitope [68]. The tool predicts the immunogenicity of a given peptide-MHC

(pMHC) complex on the basis of the physiochemical properties of constituting

amino acid and their position within the peptide sequence.

Screening of Helper T lymphocyte (HTL) Epitopes. To identify the Helper T

lymphocyte epitopes from NiV proteins, the IEDB tool "MHC-II Binding

Predictions" (http://tools.iedb.org/mhcii/) was used. Peptides with IC50 values

<50 nM are considered high affinity, <500 nM intermediate affinity and <5000 nM

low affinity [69-72]. The tool generates percentile rank for each potential peptide.

This percentile rank is generated by the combination of three different methods

viz. combinatorial library, SMM_align and Sturniolo and by comparing the score

of the peptide against the scores of other random five million 15-mer peptides of

SWISSPROT database [69-72]. The rank from the consensus of all three

methods was generated by the median percentile rank of the three methods. The

lower the value of percentile the higher would be the rank.

Population Coverage by CTL and HTL epitopes. The "Population Coverage" tool

of IEDB (http://tools.iedb.org/population/) was used to elucidate the world human

population coverage by the shortlisted 33 CTL and 38 HTL epitopes derived from

nine NiV proteins [73]. The T cells recognize complex between a specific major

MHC molecule and a particular pathogen-derived epitope. The given epitope will

elicit a response only in an individual that express an MHC molecule, which is

capable of binding that particular epitope. This denominated MHC restriction of T

cell responses and the MHC polymorphism provide basis for population coverage

study. The MHC types are expressed at dramatically different frequencies in

different ethnicities. Hence, a vaccine with optimized population coverage could

be of greater relevance in fighting NiV disease [73]. Clinical administration of

multiple-epitopes involving both the CTL and the HTL epitopes are predicted

here to have a greater probability of larger human population coverage

worldwide.

B Cell Epitope Prediction

Sequence-based B Cell epitope prediction. Protein sequence-based six different

methods were utilized to screen linear B cell epitopes from nine different NiV

proteins. These methods are available at "B Cell Epitope Prediction Tools" of

IEDB server (http://tools.iedb.org/bcell/). In this screening the parameters such

as hydrophilicity, flexibility, accessibility, turns, exposed surface, polarity and

antigenic propensity of the polypeptides are correlated with their localization in

the protein. This allows the search for continuous epitopes prediction from

protein sequence. The prediction is based on the propensity scales for each of

the 20 amino acids. For a window size n, the i - (n-1)/2 neighboring residues on

each side of residue i are used to compute the score for the residue i [74-79].

These methods utilize the propensity scale method as well as the physiochemical

properties of the given antigenic sequence to screen potential epitopes using

"Bepipred Linear Epitope Prediction", "Chou & Fasman Beta-Turn Prediction",

"Emini Surface Accessibility Prediction", "Karplus & Schulz Flexibility Prediction",

"Kolaskar & Tongaonkar Antigenicity" and "Parker Hydrophilicity Prediction" tools

[74-79].

Structure-based B cell epitope prediction. The Ellipro (ElliPro: Antibody Epitope

Prediction tool; http://tools.iedb.org/ellipro/) and the DiscoTope2.0 (DiscoTope:

Structure-based   Antibody   Prediction   tool;   http://tools.iedb.org/discotope/)

methods available at IEDB, were used to screen the linear and the discontinuous

B cell epitopes [80, 81]. The ElliPro method analyses on the basis of the location

of residue in the protein's 3D structure. The residues lying outside of an ellipsoid

covering 90% of the inner core protein residues score highest Protrusion Index

(PI) of 0.9; and so on. The discontinuous epitopes predicted by the ElliPro tool

are clustered on the basis of the distance "R" in Å between two residue's centers

of mass lying outside of the largest possible ellipsoid. The larger value of R

indicates larger distant residues (residue discontinuity) are screened in the

epitopes. The Discotope 2.0 method is based on the "contact number" of the

residue's Cα carbon atom as well as on the propensity of a given residue to be a

part of an epitope [80, 81]. The residue "contact number" is the number of Cα

atoms in the antigen within a distance of 10 Å of the particular residue's Cα atom.

A low contact number would indicate the residue being close to the surface or in

protruding regions of the antigen's structures.

Characterisation of potential epitopes

Epitope conservation analysis. The shortlisted CTL, HTL and B cell epitopes

screened from nine NiV proteins were analysed for the conservancy of their

amino   acid   sequence   by   "Epitope   Conservancy   Analysis"   tool

(http://tools.iedb.org/conservancy/) of IEDB. The epitope conservancy is the

fraction of protein sequences that contain that particular epitope. The analysis

was done against their entire respective source protein sequences of NiV

proteins retrieved from the NCBI protein database [82].

Epitope   Toxicity   prediction.   The   tool   ToxinPred

(http://crdd.osdd.net/raghava/toxinpred/multi_submit.php) was used to analyse

the toxicity of shortlisted CTL, HTL and B cell epitopes. The tool allows to

identifying highly toxic or non-toxic short peptides. The toxicity check analysis

was done by the "SVM (Swiss-Prot) based" (support vector machine) method

utilizing dataset of 1805 sequences as positive, 3593 negative sequences from

Swissprot as well as an alternative dataset comprises the same 1805 positive

sequences and 12541 negative sequences from TrEMBLE [83].

Overlapping residue analysis. The overlapping residue analysis for the shortlisted

CTL, HTL and the B cell linear epitopes was performed by the Multiple Sequence

Alignment   (MSA)   analysis   by   Clustal   Omega   tool

(https://www.ebi.ac.uk/Tools/msa/clustalo/)   of   EBI   (European   Bioinformatics

Institute) [84]. The Clustal Omega multiple sequence alignment tool virtually

aligns any number of protein sequences and delivers an accurate alignment.

Epitope selected for molecular interaction study with HLA allele and TAP

transporter. On the basis of the overlapping residue analysis of shortlisted CTL,

HTL and linear B cell epitopes few numbers of CTL and HTL epitopes were

chosen for further analysis involving stable interaction with their respective HLA

allele binders (homology models) and TAP cavity interaction (Supplementary

table S2, S3 & S4, Figure 2). These epitopes were chosen on the basis of them

having partial or full overlapping sequence region amongst all three types of

epitopes (CTL, HTL and B Cell), or having full sequence overlap amongst any of

the two types of epitopes, or having the highest number of the HLA allele

binders.

Molecular interaction analysis of selected epitopes with HLA allele and TAP

transporter.

Tertiary structure modelling of HLA alleles and selected T cell epitopes. The

Swiss-model was used for homology modelling of the HLA class I and II allele

binders of shortlisted epitopes [64]. The amino acid sequence of the HLA allele

binders were retrieved from Immuno Polymorphism Database (IPD-IMGT/HLA)

(https://www.ebi.ac.uk/ipd/imgt/hla/allele.html).   Templates   for   homology

modelling were chosen on the basis of highest amino acid sequence similarity.

All the HLA allele models were further validated by their QMEAN value. The

QMEAN value gives a composite quality estimate involving both global as well as

local analysis of the model [85]. Generated models having acceptable QMEAN

value (cutoff -4.0) were chosen for further studies (Supplementary table S2).

The "Natural Peptides Module for Beginners" module of PEPstrMOD

(http://osddlinux.osdd.net/raghava/pepstrmod/nat_ss.php)   was   utilized   to

generate tertiary structures for the selected CTL and HTL epitopes [86]. The time

window for simulation was set to 100 picoseconds (ps) in a vacuum environment.

Molecular interaction analysis of selected CTL and HTL epitopes with HLA

alleles. The AutoDock 4.2 (ADT) and the AutoDock Vina were used for *in-silico*

molecular docking study of the selected CTL and HTL epitopes with their

respective HLA class I and II allele binders [87, 88]. The generated docked

complexes were studied for their stable nature by molecular dynamics simulation.

MD simulation was performed by the Gromacs 5.1.4 using the Optimized

Potential for Liquid Simulations - all-atom force field (OPLS-AA) [89, 90].

Molecular interaction analysis of selected CTL epitopes with TAP transporter.

TAP transporter plays an important role in the presentation of CTL epitopes.

Following the immuno-proteasomal processing of xenobiotic proteins, the

fragmented peptides get transported to the endoplasmic reticulum (ER) through

the TAP transporter. Finally these short peptides reach to the Golgi apparatus

from which they get presented on the cell surface [91]. Molecular interaction

study of the shortlisted CTL epitopes with the TAP transporter cavity was

performed by molecular docking using AutoDock Vina [87, 88]. As structural

model the cryo-EM structure of TAP transporter (PDB ID: 5u1d) after removing

the antigen from TAP cavity of the original structure [56] was used for epitope-

TAP interaction study.

Characterisation of designed Multi-Epitope Vaccines

Interferon gamma inducing epitope prediction. From the designed amino acid

sequence of both the MEVs potential interferon gamma (IFN-γ) epitopes were

screened   via   the   "IFNepitope"   server

(http://crdd.osdd.net/raghava/ifnepitope/scan.php) using the "Motif and SVM

hybrid", (MERCI: Motif-EmeRging and with Classes-Identification, and SVM:

support vector machine) method. The tool predicts peptides from protein

sequences having the capacity to induce IFN-gamma release from CD4+T cells.

This module generates overlapping IFN-gamma inducing peptides from query

sequence. For the screening, IEDB database with 3705 IFN-gamma inducing

and 6728 non-inducing MHC class II binders is utilized [92, 93].

MEVs allergenicity and antigenicity prediction. The designed MEVs were further

analysed for allergenicity and antigenicity prediction by utilizing the AlgPred

(http://crdd.osdd.net/raghava/algpred/submission.html)   and   the   Vaxigen

(http://www.ddg-pharmfac.net/vaxijen/VaxiJen/VaxiJen.html)   tools   respectively

[94, 95]. The AlgPred prediction compares the sequence similarity between

known epitopes with any region of the submitted protein. For the screening of

allergenicity, the Swiss-prot dataset consisting of 101725 non-allergens and 323

allergens is utilized [94]. The VaxiJen performs an alignment-free approach,

solely based on comparing the physicochemical properties of the query amino

acid sequence with the Swiss-prot dataset. For prediction of antigenicity, the

Bacterial, viral and the tumour protein datasets were used to derive models for

the prediction of whole protein antigenicity. Every set consisted of known 100

antigens and 100 non-antigens [95].

Physicochemical property analysis of designed MEVs. The ProtParam

(https://web.expasy.org/protparam/)   tool   was   utilized   to   analyse   the

physiochemical properties of the designed CTL and HTL MEVs [96]. The

ProtParam analysis performs an empirical investigation for the given query amino

acid sequence. ProtParam computes various physicochemical properties derived

from a given protein sequence.

Tertiary structure modelling and refinement of MEVs. The tertiary structure of the

designed CTL and HTL MEVs were calculated by homology modelling utilizing

the   RaptorX   structure   prediction   tool

(http://raptorx.uchicago.edu/StructurePrediction/predict/)   [97].   RaptorX   predicts

template-based secondary and tertiary structures, protein-protein contacts,

solvent accessibility, disordered regions and binding sites for given protein

sequence, having close or even less similar homologs in the Protein Data Bank

(PDB). RaptorX also assigns confidence scores to indicate the quality of a

predicted 3D model [98-100]. Quality assessment for both the generated

homology models of CTL and HTL MEVs was performed by their respective P-

values. The P-value for a predicted homology model is a probability score of the

generated model being worse than the best. Hence the P-value indicates a

relative quality of the generated model in terms of modelling error, combining the

global distance test (GDT) and the un-normalized global distance test (uGDT)

indicating the error involved at each residue. The smaller the P-value the greater

the quality of a predicted model.

The refinement of both the generated MEV models was performed by

ModRefiner (https://zhanglab.ccmb.med.umich.edu/ModRefiner/) and

GalaxyRefine tool (http://galaxy.seoklab.org/cgi-bin/submit.cgi?type=REFINE)

[101, 102]. Modrefiner improves the physical realism and structural accuracy of

the model using a two-step atomic-level energy minimization. ModRefiner is an

algorithm for the atomic-level, high-resolution protein tertiary structure

refinement. Both the side-chain and the backbone atoms of a used protein

structure are flexible during the structure refinement simulations. The

conformational search is guided by a composite of physics and knowledge based

force field. The tool generates significant improvement in the physical quality of

the local structures [101]. TM-score generated by ModRefiner indicates the

structural similarity of the refined model with the original input model. Closer the

TM-Score to 1, higher would be the similarity of original and the refined model.

The GalaxyRefine tool refines the query tertiary structure by repeated structure

perturbation as well as by utilizing the subsequent structural relaxation by the

molecular dynamics simulation. The tool GalaxyRefine generates reliable core

structures from multiple templates and then re-builds unreliable loops or termini

by using an optimization-based refinement method [100, 103]. To avoid any

breaks in the 3D model GalaxyRefine uses triaxial loop closure method. The

MolProbity score generated for a given refined model indicates the log-weighted

combination of the clash score (the number of atomic clashes per 1000 atoms),

the Ramachandran favored backbone torsion angles and the percentage of bad

side-chain rotamers (the percentages of rotamer outliers). The 'GDT-HA' (Global

Distance Test-High Accuracy) generated by the tool indicates the backbone

structure accuracy; 'RMSD' (Root mean Square Deviation) indicates the overall

structural deviation in refined model from the initial model and the 'Rama favored'

indicates percentage of Ramachandran favored residues.

Validation of CTL and HTL MEVs refined models. Both the refined CTL and HTL

MEV 3D models were further validated by RAMPAGE analysis tool

(http://mordred.bioc.cam.ac.uk/~rapper/rampage.php) [104, 105]. The generated

Ramachandran plots for the MEV models show the sterically allowed and

disallowed residues along with their dihedral psi (ψ) and phi (φ) angles.

Discontinuous B-cell epitope prediction from MEVs. Both the generated tertiary

models of designed CTL and HTL MEVs were subjected to discontinuous B cell

epitopes prediction. The structure-based discontinuous B cell epitopes were

screened from both the MEV models by utilizing the ElliPro method as described

earlier [80].

Molecular interaction analysis of MEVs with TLR-3

Molecular interaction analysis of both the designed MEVs with TLR-3, was

performed by molecular docking and molecular dynamics simulation. Molecular

docking   was   performed   by   PatchDock   server

(http://bioinfo3d.cs.tau.ac.il/PatchDock/) [106-108]. PatchDock utilizes algorithm

for unbound (real life) docking of molecules for protein-protein. The algorithm

carries out the rigid docking, with the surface variability/flexibility implicitly

addressed through liberal intermolecular penetration. The algorithm focuses on

the (i) initial molecular surface fitting on localized, curvature based surface

patches (ii) use of Geometric Hashing and Pose Clustering for initial

transformation detection (iii) computation of shape complementarity utilizing the

Distance Transform (iv) efficient steric clash detection and geometric fit scoring

based on a multi-resolution shape representation and (v) utilization of biological

information by focusing on hot spot rich surface patches [106-108]. For molecular

docking, the 3D structure of the human TLR-3 ectodomain (ECD) was retrieved

from the PDB databank (PDB ID: 2A0Z). Further, the molecular dynamics

simulation study of the MEVs-TLR-3 complexes was performed by YASARA tool

(Yet Another Scientific Artificial Reality Application) [109]. The MD simulations

studies were carried out in an explicit water environment in a dodecahedron

simulation box at a stabilized temperature of 298K, the pressure of 1atm and pH

7.4, with periodic cell boundary condition. The solvated systems were neutralized

with counter ions (NaCl) (concentration 0.9 M). The AMBER14 force field was

used on the systems during MD simulation [110]. The Long-range electrostatic

energy and forces were calculated using particle-mesh-based Ewald method

[111]. The solvated structures were energy minimized by the steepest descent

method at a temperature of 298K and a stable pressure of 1 atm. Further, the

complexes were equilibrated for period of 1 ns. After equilibration, a production

MD simulation was run for 100 ns at a stable temperature and pressure and time

frames were saved at every 100 ps, for each MD simulations. The RMSD and

RMSF values for Cα, Backbone and all the atoms of all the CTL and HTL MPVs

in complex with TLR3 were analyzed for each MD simulation study.

***In-silico* analysis of MEVs for cloning and expression**

Analysis of cDNA of the MEVs for cloning and expression in mammalian cell line.

Complementary DNA of both the MEVs, codon optimized for expression in

mammalian cells were generated by Java Codon Adaptation Tool

(http://www.jcat.de/). The generated cDNA was further analysed with the

GenScript Rare Codon Analysis Tool (https://www.genscript.com/tools/rare-

codon-analysis) which calculates the GC content, Codon Adaptation Index (CAI)

and the Tandem rare codon frequency for a given cDNA [112-114]. The CAI

indicates the possibility of cDNA expression in a chosen expression system. The

tandem rare codon frequency indicates the presence of low-frequency codons in

the given cDNA.

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AUTHOR CONTRIBUTION:

Idea conceived, methodology designed and performed by S.S., MD simulation done by S.V., critical data analysis and scientific writing was done by S.S., M.K. and M.Kolbe, draft was finalized by S.S., S.V., M.K., M.Kolbe, A.K.S. and K.C.P.

ADDITIONAL INFORMATION:

Authors declare to have no competing interests.

Supplementary figure S1. Workflow chart.

Supplementary figure S2. (A) Rg (radius of gyration) for the CTL epitope – HLA class I allele complexes, across the time window of 1 nano second. **(B)** Rg for the HTL epitope – HLA class II allele complexes, across the time window of 1 nano second.

Supplementary figure S3. (A) RMS fluctuation in nanometers for all the atoms of the CTL epitope – HLA class I allele complexes. **(B)** RMS fluctuation in nanometers for all the atoms of the HTL epitope – HLA class II allele complexes.

Supplementary figure S4. (A) B-Factor of CTL epitope – HLA class I allele complexes **(B)** B-Factor of HTL epitope – HLA class II allele complexes. Epitopes are shown in sticks and HLA alleles are shown in gray cartoons. B-factor is indicated by rainbow (VIBGYOR) colour, blue for stable region and red for most unstable region.

914 **Supplementary table S1. Protein sequence retrieval, tertiary structures**
915 **retrieval and homology modeling of nine Nipah proteins.** Nipah protein
916 sequences were retrieved from NCBI. Available structure files (pdb) for Nipah
917 proteins were retrieved from RCSB PDB. Nipah proteins with no tertiary structure
918 available were subjected to homology modeling by Swissmodel.
919

920 **Supplementary table S2. Homology modeling for HLA alleles.** Tertiary
921 structure of HLA alleles were modeled by homology modeling using SwissModel
922 server. Templates were chosen with highest sequence identity. Generated
923 models with acceptable QMEAN value were chosen for further studies.
924

Supplementary table S3. Shortlisted high scoring CTL epitopes. Selected high
scoring CTL epitopes and their respective HLA alleles binders are listed. *In-silico*
analysis have shown all the selected epitopes to be non-toxic (Non-Toxin) as well as
they show significant conservancy. ToxinPred analysis is based on the ToxinPred main
dataset used by "ToxinPred" algorithm to predict toxicity of any unknown peptide. #
Epitope match with previous studies indicating consensus in epitope screening by
different approaches and methods.
932

933 **Supplementary table S4. Shortlisted high scoring HTL epitopes.** Selected
934 high scoring HTL epitopes and their respective HLA alleles binders are listed
935 above. *In-silico* analysis have shown all the selected epitopes to be non-toxic
936 (Non-Toxin) as well as they show significant conservancy.
937

938 **Supplementary table S5. World population coverage by the shortlisted CTL**
939 **and HTL epitopes combined.** The cumulative percent of world (countries as
940 listed in table) population coverage is 97.88% by the joint administration of
941 selected CTL and HTL epitopes as vaccine candidate.
942

Supplementary table S6. Shortlisted B Cell epitopes. BepiPred Linear B Cell
epitopes showing sequence overlap with CTL and HTL epitopes are shortlisted
above. *In-silico* analysis have shown all the selected epitopes to be non-toxic
(Non-Toxin) as well as they show significant amino acid sequence conservancy.
#Epitope match with previous studies indicating consensus in epitope screening by
different approaches and methods.
949

950 **Supplementary table S7. CTL epitope prediction.** Detailed scoring of all
951 screened CTL epitopes and their respective HLA class I allele binders. CTL
952 epitopes were chosen on the basis of high "Total score" and higher number of
953 HLA allele binders. Total score is a combined score of TAP score, MHC score,
954 Proteasome score and Processing score.
955

956 **Supplementary table S8. HTL epitope prediction.** Percentile rank of HTL
957 epitopes and their respective HLA class II allele binders. HTL epitopes were
958 screened on the basis of percentile rank (lower the percentile number, higher the

rank) and larger number of HLA allele binders. Last column show the method used for epitope screening.

Supplementary table S9. INF-γ epitopes from CTL and HTL MEVs. INF-γ inducing (POSITIVE) epitopes from CTL and HTL MEVs were screened by using "Motif and SVM hybrid" (MERCI & SVM) approaches.

Supplementary table S10. Refinement models of CTL and HTL MEVs. CTL and HTL MEVs models were refined by GalaxyWEB server and used for further studies. After refinement in particular Rama favored residues increased significantly.

Supplementary table S11. B Cell discontineous epitopes of CTL & HTL MEVs. Discontinuous B Cell epitopes predicted by ElliPro (IEDB) from CTL & HTL MEVs.

Protein	Overlapping regions of predicted B cell Bepipred linear Epitope and epitopes predicted by other protein sequence and structure based methods	Overlapping regions of sequence based B cell linear epitope prediction					Overlapping regions of structure based B cell epitope prediction		
		Chou & Fasman Beta-Turn	Emini Surface Accessibility	Karplus & Schulz Flexibility	Kolaskar & Tongaonkar Antigenicity	Parker Hydrophilicity	DiscoTope: discontinuous epitopes	ElliPro Linear	Discontinuous
Fusion Protein	215 GPNLQDPVSNSM 226	215-226		215-226	215-216	217-226			
Glycoprotein	271 WTPNPNT 278	271-278	271-278	272-278		272-278			271-278
Glycoprotein	459 SWDTMI 464	459-460							463-464
Glycoprotein	529 NQTAE 533	529-532	529-533	529-533		529-533			
Matrix Protein	184 SGIYM 188	184-188		184-188		184-188			
Matrix Protein	330 SIPREFMIY 338	330-335		330-336		331-334	332-333	332-333	333-338
Matrix Protein	340 DVFIDNTGRI 349	341-349		340-349		340-344			340-349
Phosphoprotein	125 GYGFTSSPERGWSDYTSGA 143	125-143	130-136	128-143		127-143			
Phosphoprotein	163 IAVSKEDR 170	166-170	166-170	163-170	163-166	165-170			
Polymerase	346 NIDN 349	348-349							
Polymerase	364 GHPILE 369	364-367			365-369				366-369
Polymerase	1095 DKSFDEDLEL 1104	1096-1103	1096-1101	1095-1104		1096-1103			1095-1104
Polymerase	1108 LMDR 1111								1108-1111
Polymerase	1409 LRLETDDYNG 1418	1412-1418	1410-1416	1412-1418		1411-1418	1409-1417	1409-1417	
Polymerase	1982 GFPIS 1986	1982-1986		1982-1986	1982-1986				1182-1986
Polymerase	2004 PVYSNPD 2010	2004-2010	2006-2010	2007-2010	2004-2008	2007-2010			2004-2010

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Figure 1. Overlapping regions amongst the linear B cell epitopes predicted

by the BepiPred method and other seven different B cell epitopes

prediction methods. B cell epitopes predicted by the BepiPred method and

other different protein sequence based (Chou.., Emini.., Karplus.., Kolaskar..,

and Parker..) and protein structure based (DiscoTope and ElliPro) prediction

methods were found to have significant consensus. Consensus overlapping

regions of BepiPred epitopes are underlined by the different colour,

corresponding to respective prediction method.

Clusters of overlapping CTL, HTL & B cell epitopes

Figure 2. Overlapping CTL, HTL and B cell epitopes. Multiple sequence

alignment performed by Clustal Omega at EBI to identify the consensus

overlapping regions of CTL (red), HTL (blue) and B cell epitopes (green)

amongst shortlisted epitopes. Epitopes with overlapping region amongst all the

three types of epitopes (CTL, HTL and B Cell epitopes), epitopes with full

sequence overlap and epitopes with the highest number of HLA allele binders

were chosen for further studies (encircled).

Figure 3 (A). Molecular Docking analysis of CTL epitopes and HLA alleles.

Molecular docking of selected CTL epitopes (cyan sticks) with their respective

HLA class I allele binders (gray sticks). The study shows the docked complexes

to have significantly negative binding energy along with hydrogen bonds (green

dots) formation in the complex interface. **(B) Molecular Docking analysis of**

HTL epitopes and HLA alleles. Molecular docking of selected HTL epitopes

(cyan sticks) with their respective HLA class II allele binders (gray sticks). The

study shows the docked complexes to have significantly negative binding energy

along with hydrogen bonds (green dots) formation in the complex interface. (*)

Indicates binding energy, shown in kcal/mol.

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Figure 4 (A). Molecular Dynamics simulation analysis of CTL epitopes and

HLA allele complexes. Molecular Dynamics simulation study reveals a stable

nature of the CTL-HLA allele complexes throughout 0.5-1 ns time window with

reasonably invariable RMSD. **(B) Molecular Dynamics simulation analysis of**

CTL epitopes and HLA allele complexes. Molecular Dynamics simulation study

reveals a stable nature of the HTL-HLA allele complexes throughout 0.5-1 ns

time window with reasonably invariable RMSD.

Figure 5. Molecular docking analysis of CTL epitopes within the TAP transporter cavity. Molecular interaction of CTL epitopes (cyan sticks) within the TAP cavity (gray ribbon/sticks) is shown in detail. For every panel of epitope-TAP complex, **(A)** shows the binding of epitope at two different sites within TAP cavity, **(B)** and **(C)** show detailed molecular interaction between epitopes and TAP cavity; (a, b) show chain A and B of TAP transporter. H-bonds are shown in yellow dots. (*) Indicates binding energy, shown in kcal/mol.

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A. CTL epitopes
B. HTL epitopes

ß Defensin 2 — GIGDPVTCLKSGAICHPVFCPRRYKQIGTCGLPGTKCCKKP

B. HTL epitopes	A. CTL epitopes
EAAAK	EAAAK
C Protein — VQMTYNWTQWLQTLY	MMASILLTLF
GGGGS	GGGGS
DLALSKYLSDLLFVF	AQITAGVALY
GGGGS	
LALSKYLSDLLFVFG	FALSNGVLF
GGGGS	
ALSKYLSDLLFVFGP	KYLSDLLFVF
Fusion Protein — LSKYLSDLLFVFGPN	GGGGS
GGGGS	MTIQAISQAF
SKYLSDLLFVFGPNL	GGGGS
GGGGS	AENPVFTVF
KYLSDLLFVFGPNLQ	GGGGS
GGGGS	AVYNNEFYY
YLSDLLFVFGPNLQD	GGGGS
GGGGS	LAMDEGYFAY
ASFSWDTMIKFGDVL	GGGGS
GGGGS	TVYHCSAVY
FSWDTMIKFGDVLTV	GGGGS
GGGGS	NYMYLICYGF
GVYNDAFLIDRINWI	GGGGS
Glycoprotein — NDAFLIDRINWISAG	YMIPRTMLEF
GGGGS	GGGGS
DAFLIDRINWISAGV	EIISDIGNY
GGGGS	GGGGS
AFLIDRINWISAGVF	TPFVDSRAY
GGGGS	GGGGS
FLIDRINWISAGVFL	YPALALNEF
GGGGS	GGGGS
Matrix Protein — IPREFMIYDDVFIDN	LDPVVTDVVY*
GGGGS	GGGGS
FMIYDDVFIDNTGRI	LVSDAKMLSY*
GGGGS	GGGGS
LSSDQVAELAAAVQE	MPSDDFSNTF
GGGGS	GGGGS
SSDQVAELAAAVQET	VSDAKMLSY*
Nucleocapsid — SDQVAELAAAVQETS	GGGGS
GGGGS	AEFFSFFRTF
DQVAELAAAVQETSA	GGGGS
GGGGS	ETDDYNGIY
QVAELAAAVQETSAG	GGGGS
GGGGS	FPISRLFNMY
*NNGNVCLVSDAKMLS	GGGGS
GGGGS	FPVMGNRIY
*NGNVCLVSDAKMLSY	GGGGS
Phosphoprotein — *GNVCLVSDAKMLSYA	IATVYTWAY
GGGGS	GGGGS
*NVCLVSDAKMLSYAP	IMKKSFKAY
GGGGS	GGGGS
*VCLVSDAKMLSYAPE	IPFLFLSAY
GGGGS	GGGGS
NIDNIHLLAEFFSFF	KWYECFLFWF
GGGGS	GGGGS
IDNIHLLAEFFSFFR	KYYQIDQPFF
GGGGS	GGGGS
DNIHLLAEFFSFFRT	LETDDYNGIY
GGGGS	GGGGS
NIHLLAEFFSFFRTF	RLFNMYRSY
GGGGS	GGGGS
IHLLAEFFSFFRTFG	SQNLLVTSY
Polymerase — LELASFLMDRRVILP	GGGGS
GGGGS	SYFGLVLVCF
ELASFLMDRRVILPR	GGGGS
GGGGS	TSDLDFVIFY
LASFLMDRRVILPRA	GGGGS
GGGGS	YPECNNILF
ASFLMDRRVILPRAA	EAAAK
GGGGS	
LDFVIFYASLTYLRR	
GGGGS	
FVIFYASLTYLRRGI	
EAAAK	

ß Defensin 3 — GIINTLQKYYCRVRGGRCAVLSCLPKEEQIGKCSTRGRKCCRRKK
HHHHHH   HHHHHH

Figure 6. Design of Multi-Epitope Vaccine (MEVs). (A) CTL and **(B)** HTL

epitopes were linked by the short peptide linker 'GGGGS'. Human β Defensin 2

and β Defensin 3 were used as an adjuvant at the N and C terminals

respectively. The short peptide EAAAK was used to link the β Defensin 2 and β

Defensin 3. Epitopes from different proteins were coloured in different colours. C

terminal 6xHis is designed as His tag. * Epitopes common to Phosphoprotein, V

Protein and W protein.

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CTL multi-epitope vaccine:

HTL multi-epitope vaccine:

Figure 7. Tertiary structure modelling of CTL and HTL Multi-Epitope

Vaccines. (A) & (F): Tertiary structural models of CTL and HTL MEVs showing

epitopes in different colours corresponds to as in Fig. 1. **(B) & (G)**: Show the

different domains of CTL and HTL MEVs. **(C) & (H)**: The overlapping linear B cell

epitope region present in CTL and HTL MEVs, shown by spheres. **(D) & (I):** From

the CTL and HTL MEVs, the INF-γ inducing epitopes are shown in cyan,

discontinuous B Cell epitopes are shown in magenta and the region common

amongst INF-γ and discontinuous B Cell epitopes are shown in wheat colour. **(E)**

& (J): RAMPAGE analysis of the refined CTL and HTL MEV models.

Figure 8. Molecular Docking and dynamics simulation study of CTL and

HTL MEVs with TLR-3. (A) CTL and **(E)** HTL MEVs (VIBGYOR) docked

complex with TLR-3 (gray). Both the complexes are forming several hydrogen

bonds in the MEV and TLR-3 interface, as shown by green dots. B-Factor of the

docked MEVs is shown by a rainbow (VIBGYOR) presentation. The regions in

blue being indicated stable and the region in red indicate unstable. In the above

complexes, most of the region of docked MEVs is in blue and with the very small

region is green, yellow or orange, hence the complexes are predicted to be very

stable. **(B)** and **(F),** RMSD as generated by the molecular Dynamics simulation

study of CTL, HTL MEVs and TLR-3 complexes. **(C)** & **(G)** Rg (radius of gyration)

across the time window of 100 nanosecond. **(D)** & **(H)**, RMS fluctuation for all the

atoms of the CTL, HTL MEVs and TLR-3 complexes.

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1958
1959
1960

Overlapping regions of predicted B cell Bepipred linear Epitope and epitopes predicted by other protein sequence and structure based methods	Overlapping regions of sequence based B cell linear epitope prediction					Overlapping regions of structure based B cell epitope prediction		
	Chou & Fasman Beta-Turn	Emini Surface Accessibility	Karplus & Schulz Flexibility	Kolaskar & Tongaonkar Antigenicity	Parker Hydrophilicity	DiscoTope: discontinuous epitopes	ElliPro Linear	ElliPro Discontinuous
Fusion Protein — 215 GPNLQDPVSNSM 226	215-226		215-226	215-216	217-226			
Glycoprotein — 271 WTPPNPNT 278	271-278	271-278	272-278		272-278			271-278
Glycoprotein — 459 SWDTMI 464	459-460							463-464
Glycoprotein — 529 NQTAE 533	529-532	529-533	529-533		529-533			
Matrix Protein — 184 SGIYM 188	184-188		184-188		184-188			
Matrix Protein — 330 SIPREFMIY 338	330-335		330-336		331-334	332-333	332-333	333-338
Matrix Protein — 340 DVFIDNTGRI 349	341-349		340-349		340-344			340-349
Phosphoprotein — 125 GYGFTSSPERGWSDYTSGA 143	125-143	130-136	128-143		127-143			
Phosphoprotein — 163 IAVSKEDR 170	166-170	166-170	163-170	163-166	165-170			
Polymerase — 346 NIDN 349	348-349							
Polymerase — 364 GHPILE 369	364-367			365-369				366-369
Polymerase — 1095 DKSFDEDLEL 1104	1096-1103	1096-1101	1095-1104		1096-1103			1095-1104
Polymerase — 1108 LMDR 1111								1108-1111
Polymerase — 1409 LRLETDDYNG 1418	1412-1418	1410-1416	1412-1418		1411-1418	1409-1417	1409-1417	
Polymerase — 1982 GFPIS 1986	1982-1986		1982-1986	1982-1986				1182-1986
Polymerase — 2004 PVYSNPD 2010	2004-2010	2006-2010	2007-2010	2004-2008	2007-2010			2004-2010

Clusters of overlapping CTL, HTL & B cell epitopes

Fusion Protein

```
                                    -------------------MTIQAISQAF-
     ----------KYLSDLLFVF----------------------
           -----YLSDLLFVFGPNLQD-------------
           -----KYLSDLLFVFGPNLQ------------
           -----SKYLSDLLFVFGPNL-----------
           ----LSKYLSDLLFVFGPN------------
           ---ALSKYLSDLLFVFGP--------------
           ----LALSKYLSDLLFVFG-------------
           -DLALSKYLSDLLFVF----------------
                          -----GPNLQDPVSNSM----------
199-236 LDLALSKYLSDLLFVFGPNLQDPVSNSMTIQAISQAFG
```

Glycoprotein

```
                                                    -------------AENPVFTVF-
                                        ------FLIDRINWISAGVFL---------------
                                        -----AFLIDRINWISAGVF--------------
                                        ----DAFLIDRINWISAGV---------------
          -----TVYHCSAVY-------          ----FSWDTMIKFGDVLTV--        ---NDAFLIDRINWISAG----------
          -------------AVYNNEFYY-        --ASFSWDTMIKFGDVL----        -GVYNDAFLIDRINWI-------------
       -WTPPNPNT--------------          -----SWDTMI------                    ----------------NQTAE---------
270-293 VWTPPNPNTVYHCSAVYNNEFYYV   454-474 YQASFSWDTMIKFGDVLTVNP   505-541 EGVYNDAFLIDRINWISAGVFLDSNQTAENPVFTVFK
```

Matrix Protein

```
                                    -----FMIYDDVFIDNTGRI-
                                    -IPREFMIYDDVFIDN-------
          ----YMIPRTMLEF-            SIPREFMIY-----DVFIDNTGRI-
          -SGIYM---------            SIPREFMIY------
183-197 DSGIYMIPRTMLEFR   330-350 SIPREFMIYDDVFIDNTGRIL
```

Nucleocapsid

```
          ------QVAELAAAVQETSAG--
          -----DQVAELAAAVQETSA-
          -----SDQVAELAAAVQETS-
          ---SSDQVAELAAAVQET--
          --LSSDQVAELAAAVQE------
382-404 LGLSSDQVAELAAAVQETSAGRQ
```

Phosphoprotein

```
                              --------LVSDAKMLSY------------
                              ---------VSDAKMLSY-----------
                              ----VCLVSDAKMLSYAPE---------
                              -----NVCLVSDAKMLSYAP--------
                              ----GNVCLVSDAKMLSYA--------
                              ----NGNVCLVSDAKMLSY--------
                              ----NNGNVCLVSDAKMLS--------
          --GYGFTSSPERGWSDYTSGA-----------------IAVSKEDR--
123-172 CTGYGFTSSPERGWSDYTSGANNGNVCLVSDAKMLSYAPEIAVSKEDRET
```

Polymerase

```
          --------AEFFSFFRTF------
          -----IHLLAEFFSFFRTFG------
          ----NIHLLAEFFSFFRT-------        ----------ASFLMDRRVILPRAA-----
          ---DNIHLLAEFFSFFRT------         ----------LASFLMDRRVILPRA-----
          --IDNIHLLAEFFSFFR------          ------ELASFLMDRRVILPR------
          -NIDNIHLLAEFFSFF-------          ----LELASFLMDRRVILP------            ----ETDDYNGIY--
          -NIDN----------------           -------------LMDR------------       -----LETDDYNGIY--
                         GHPILE-          --DKSFDEDLEL------                  -LRLETDDYNG----
345-370 LNIDNIHLLAEFFSFFRTFGHPILEA   1093-1124 FHDKSFDEDLELASFLMDRRVILPRAAHEILD   1408-1422 RLRLETDDYNGIYHL
```

Polymerase

```
                                    ---------SYFGLVLVCF----------
                                    --------RLFNMYRSY-----------
          -TSDLDFVIFY----------     --FPISRLFNMY-------------------
          ---VIFYASLTYLRRGI-                                  ------PVYSNPD--
          ----LDFVIFYASLTYLRR---    -GFPIS--------------------
1715-1736 NTSDLDFVIFYASLTYLRRGII   1981-2012 PGFPISRLFNMYRSYFGLVLVCFPVYSNPDST
```

A. CTL epitopes - HLA Class I allele complexes

*-5.9
1. AEFFSFFRTF-B4403

*-5.1
2. AENPVFTVF-B4402

*-4.1
3. AENPVFTVF-B4403

*-7.3
4. AVYNNEFYY-A1101

*-6.3
5. AVYNNEFYY-A3002

*-6.4
6. KYLSDLLFVF-A2301

*-5.8
7. KYLSDLLFVF-A2402

*-4.7
8. LDPVVTDVVY-B3501

*-5.3
9. RLFNMYRSY-A3002

*-6.2
10. RLFNMYRSY-A3201

*-5.8
11. RLFNMYRSY-B1501

*-4.2
12. VSDAKMLSY-A0101

*-5.3
5. GNVCLVSDAKMLSYA-DRB1-0301

*-6.4
6. IDNIHLLAEFFSFFR-DPA1-0103

*-5.7
7. IDNIHLLAEFFSFFR-DPB1-0201

*-6.8
8. IHLLAEFFSFFRTFG-DPA1-0103

*-6.1
9. IHLLAEFFSFFRTFG-DPB1-0201

*-6.2
10. KYLSDLLFVFGPNLQ-DRB3-0101

*-5.3
11. LALSKYLSDLLFVFG-DRB3-0101

*-6.1
12. LSKYLSDLLFVFGPN-DRB3-0101

*-5.2
13. NGNVCLVSDAKMLSY-DRB1-0301

*-7.5
14. NIDNIHLLAEFFSFF-DPA1-0103

*-5.8
15. NIDNIHLLAEFFSFF-DPB1-0201

*-5.9
16. NIHLLAEFFSFFRTF-DPA1-0103

*-5.7
17. NIHLLAEFFSFFRTF-DPB1-0201

*-4.0
18. NNGNVCLVSDAKMLS-DRB1-0301

*-5.6
19. NVCLVSDAKMLSYAP-DRB1-0301

*-5.7
20. SKYLSDLLFVFGPNL-DRB3-0101

B. HTL epitopes - HLA Class II allele complexes

*-5.3
1. ALSKYLSDLLFVFGP-DRB3-0101

*-5.6
2. DLALSKYLSDLLFVF-DRB3-0101

*-5.5
3. DNIHLLAEFFSFFRT-DPA1-0103

*-6.2
4. DNIHLLAEFFSFFRT-DPB1-0201

*-5.1
21. VCLVSDAKMLSYAPE-DRB1-0301

*-5.1
22. YLSDLLFVFGPNLQD-DRB3-0101

A

1. AEFFSFFRTF-B4403
2. AENPVFTVF-B4402
3. AENPVFTVF-B4403
4. AVYNNEFYY-A1101
5. AVYNNEFYY-A3002
6. KYLSDLLFVF-A2301
7. KYLSDLLFVF-A2402
8. LDPVVTDVVY-B3501
9. RLFNMYRSY-A3002
10. RLFNMYRSY-A3201
11. RLFNMYRSY-B1501
12. VSDAKMLSY-A0101

RMSD (nm)

TIME (ns)

B

1. ALSKYLSDLLFVFGP-DRB3-0101
2. DLALSKYLSDLLFVF-DRB3-0101
3. DNIHLLAEFFSFFRT-DPA1-0103
4. DNIHLLAEFFSFFRT-DPB1-0201
5. GNVCLVSDAKMLSYA-DRB1-0301
6. IDNIHLLAEFFSFFR-DPA1-0103
7. IDNIHLLAEFFSFFR-DPB1-0201
8. IHLLAEFFSFFRTFG-DPA1-0103
9. IHLLAEFFSFFRTFG-DPB1-0201
10. KYLSDLLFVFGPNLQ-DRB3-0101
11. LALSKYLSDLLFVFG-DRB3-0101
12. LSKYLSDLLFVFGPN-DRB3-0101
13. NGNVCLVSDAKMLSY-DRB1-0301
14. NIDNIHLLAEFFSFF-DPA1-0103
15. NIDNIHLLAEFFSFF-DPB1-0201
16. NIHLLAEFFSFFRTF-DPA1-0103
17. NIHLLAEFFSFFRTF-DPB1-0201
18. NNGNVCLVSDAKMLS-DRB1-0301
19. NVCLVSDAKMLSYAP-DRB1-0301
20. SKYLSDLLFVFGPNL-DRB3-0101
21. VCLVSDAKMLSYAPE-DRB1-0301
22. YLSDLLFVFGPNLQD-DRB3-0101

RMSD (nm)

TIME (ns)

1. Epitope AEFFSFFRTF through TAP cavity

2. Epitope AENPVFTVF through TAP cavity

3. Epitope AVYNNEFYY through TAP cavity

4. Epitope KYLSDLLFVF through TAP cavity

5. Epitope LDPVVTDVVY through TAP cavity

6. Epitope RLFNMYRSY through TAP cavity

7. Epitope VSDAKMLSY through TAP cavity

B. HTL epitopes | **A. CTL epitopes**

ß Defensin 2
GIGDPVTCLKSGAICHPVFCPRRYKQIGTCGLPGTKCCKKP

EAAAK | EAAAK

C Protein
VQMTYNWTQWLQTLY | MMASILLTLF

GGGGS

Fusion Protein

B. HTL epitopes	A. CTL epitopes
DLALSKYLSDLLFVF	AQITAGVALY
LALSKYLSDLLFVFG	FALSNGVLF
ALSKYLSDLLFVFGP	KYLSDLLFVF
LSKYLSDLLFVFGPN	MTIQAISQAF
SKYLSDLLFVFGPNL	AENPVFTVF
KYLSDLLFVFGPNLQ	AVYNNEFYY
YLSDLLFVFGPNLQD	LAMDEGYFAY

(linkers: GGGGS)

Glycoprotein

B. HTL epitopes	A. CTL epitopes
ASFSWDTMIKFGDVL	TVYHCSAVY
FSWDTMIKFGDVLTV	NYMYLICYGF
GVYNDAFLIDRINWI	YMIPRTMLEF
NDAFLIDRINWISAG	EIISDIGNY
DAFLIDRINWISAGV	TPFVDSRAY
AFLIDRINWISAGVF	YPALALNEF
FLIDRINWISAGVFL	LDPVVTDVVY*

(linkers: GGGGS)

Matrix Protein

B. HTL epitopes	A. CTL epitopes
IPREFMIYDDVFIDN	LVSDAKMLSY*
FMIYDDVFIDNTGRI	MPSDDFSNTF

(linkers: GGGGS)

Nucleocapsid

B. HTL epitopes	A. CTL epitopes
LSSDQVAELAAAVQE	VSDAKMLSY*
SSDQVAELAAAVQET	AEFFSFFRTF
SDQVAELAAAVQETS	ETDDYNGIY
DQVAELAAAVQETSA	FPISRLFNMY
QVAELAAAVQETSAG	FPVMGNRIY

(linkers: GGGGS)

Phosphoprotein

B. HTL epitopes	A. CTL epitopes
*NNGNVCLVSDAKMLS	IATVYTWAY
*NGNVCLVSDAKMLSY	IMKKSFKAY
*GNVCLVSDAKMLSYA	IPFLFLSAY
*NVCLVSDAKMLSYAP	KWYECFLFWF
*VCLVSDAKMLSYAPE	KYYQIDQPFF

(linkers: GGGGS)

Polymerase

B. HTL epitopes	A. CTL epitopes
NIDNIHLLAEFFSFF	LETDDYNGIY
IDNIHLLAEFFSFFR	RLFNMYRSY
DNIHLLAEFFSFFRT	SQNLLVTSY
NIHLLAEFFSFFRTF	SYFGLVLVCF
IHLLAEFFSFFRTFG	TSDLDFVIFY
LELASFLMDRRVILP	YPECNNILF
ELASFLMDRRVILPR	
LASFLMDRRVILPRA	
ASFLMDRRVILPRAA	
LDFVIFYASLTYLRR	
FVIFYASLTYLRRGI	

(linkers: GGGGS)

EAAAK | EAAAK

ß Defensin 3
GIINTLQKYYCRVRGGRCAVLSCLPKEEQIGKCSTRGRKCCRRKK

HHHHHH | HHHHHH

CTL multi-epitope vaccine:

HTL multi-epitope vaccine:

Legend:
- General/Pre-Pro/Proline Favoured
- General/Pre-Pro/Proline Allowed
- Glycine Favoured
- Glycine Allowed

A. B-factor of CTL epitope & HLA I allele complesxes

1. AEFFSFFRTF-B4403 2. AENPVFTVF-B4402 3. AENPVFTVF-B4403 4. AVYNNEFYY-A1101

5. AVYNNEFYY-A3002 6. KYLSDLLFVF-A2301 7. KYLSDLLFVF-A2402 8. LDPVVTDVVY-B3501

9. RLFNMYRSY-A3002 10. RLFNMYRSY-A3201 11. RLFNMYRSY-B1501 12. VSDAKMLSY-A0101

B. B-factor of HTL epitope & HLA II allele complesxes

1. ALSKYLSDLLFVFGP-DRB3-0101 2. DLALSKYLSDLLFVF-DRB3-0101 3. DNIHLLAEFFSFFRT-DPA1-0103 4. DNIHLLAEFFSFFRT-DPB1-0201

5. GNVCLVSDAKMLSYA-DRB1-0301 6. IDNIHLLAEFFSFFR-DPA1-0103 7. IDNIHLLAEFFSFFR-DPB1-0201 8. IHLLAEFFSFFRTFG-DPA1-0103

9. IHLLAEFFSFFRTFG-DPB1-0201 10. KYLSDLLFVFGPNLQ-DRB3-0101 11. LALSKYLSDLLFVFG-DRB3-0101 12. LSKYLSDLLFVFGPN-DRB3-0101

13. NGNVCLVSDAKMLSY-DRB1-0301 14. NIDNIHLLAEFFSFF-DPA1-0103 15. NIDNIHLLAEFFSFF-DPB1-0201 16. NIHLLAEFFSFFRTF-DPA1-0103

17. NIHLLAEFFSFFRTF-DPB1-0201 18. NNGNVCLVSDAKMLS-DRB1-0301 19. NVCLVSDAKMLSYAP-DRB1-0301 20. SKYLSDLLFVFGPNL-DRB3-0101

21. VCLVSDAKMLSYAPE-DRB1-0301 22. YLSDLLFVFGPNLQD-DRB3-0101